Fast measurement of human body posture based on three-dimensional optical information

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Abstract: The three-dimensional measurement of human body posture is of great significance to the comfort evaluation of car seat design. In order to acquire the 3D data of the human body in the car quickly and accurately, a method of 3D data acquisition based on binocular vision was adopted, which combined the structured light with the marked points, and realized the rapid reconstruction of 3D point cloud of the human body and the automatic and rapid measurement of 3D attitude (distance and angle). The experimental results show that when the distance is more than 2 m and the measuring range is $1.5 \text{ m} \times 2 \text{ m}$, the measurement accuracy of human body posture can reach 0.03 mm, which meets the demand of high-precision three-dimensional data acquisition of automobile human body posture. Compared with traditional three-dimensional measurement method, the three-dimensional automatic measurement method used in this paper not only has a high degree of automation, but also has the advantages of high accuracy, fast speed and strong robustness.

Key words:human body posture;three-dimensional measurement;structured light;marked pointsCLC number:TP391.41Document code: ADOI: 10.3788/IRLA20200079

基于三维光学信息的人体姿态快速测量

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摘 要:汽车中人体姿态三维测量对汽车座椅设计的舒适性评价具有重要的意义。为了快速准确地 获取车内人体三维数据,文中采用一种基于双目视觉的立体三维数据获取方法,该方法将结构光与标 记点相结合,实现了人体三维点云快速重建以及三维姿态(距离、角度)自动快速测量。实验结果表 明,该方法在距离2m以上,测量范围1.5m×2m时,人体姿态测量精度可以达到0.03mm,满足了汽 车人体姿态高精度三维数据采集的需求。与传统的汽车人体姿态三维测量方法相比,文中所使用的三 维自动测量方法不仅自动化程度高,而且具有测量精度高、速度快、鲁棒性强的优点。 关键词:人体姿态: 三维测量: 结构光: 标记点

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0 Introduction

The traditional human measurement methods mainly use soft ruler, altimeter, slider and other contacted measuring tools to measure the human body. Although these contact measurement methods can obtain more detailed human body data, many practical issues during measurement such as heterosexual contact, fatigue, outfitting and personal measurement skills can affect measurement accuracy or measurement experience.

As a branch of modern image measurement technology, 3D anthropometry is a 3D measurement technology based on modern optics, including fusion electronics, computer graphics, information processing, and computer vision. Compared with traditional measurement technology, 3D anthropometric measurement can measure the dimension of multiple parts of the human body in a few seconds, with high precision. Among them, non-contact optical 3D measurement greatly shortens the measurement time and improves the measurement accuracy, provides more accurate data aided design for the subsequent processing of 3D body data, and is beneficial to data management and recording^[1]. The widely used active 3D measurement methods mainly include Moire interferometry method, time-of-flight (TOF) method, laser triangulation method and structured light method. Moire interferometry^[2]uses two equally spaced fringe gratings with small angular displacements to generate moiré fringes, and then obtains the phase information of the object, based on which the depth value can be calculated through the demodulation function. Although this method can acquire the depth information of an object in real time, it is not suitable for the 3D measurement of human body. The ToF^[3]method calculates the depth information of an object by transmitting ultrasonic waves or light pulses of a certain frequency and measuring the time difference between the transmitted signal and received signal. The advantage of this method is that it is not affected by the gray level of the object or the characteristics of the object. However, the time-of-flight camera has a lower resolution and its

cost is high. The Laser Triangulation^[4] method acquires the depth information of an object by calculating the angular offset between the emitted laser beam and the received laser beam. The method can effectively avoid the influence of external light on the measurement result during the measurement process and its measurement accuracy is high. But its measurement speed is slow. The 3D non-contact measurement technology based on structured light has the advantages of fast measurement speed, large measurement range and high robustness^[5,6]. Considering the measurement accuracy, measurement range and equipment cost, we use the structured light method as the 3D reconstruction method to obtain highprecision 3D data.

Lu proposed^[7] a novel human detection method using depth information taken by the Kinect for Xbox 360. This method could measure 3D face model in real time, but the accuracy was not satisfied. Andreas^[8] compared 3D laser-based body scans (BS) with classical manual anthropometric (CA) assessments with respect to feasibility, reliability, and validity. Schwarz-Müller proposed a newly developed 'positioning aid'^[9] that stabilized the posture during the scanning process and was invisible on scans which overall increased the precision of the software-assisted extraction of body dimensions.

Car seat comfort is one of the most important indicators for car manufacturers and consumers, and has a relatively high impact on human health. Traditionally based on the 2D image skeleton extraction method, it is difficult to obtain the 3D spatial information of the human posture. The 3D measurement method based on optical motion capture technology has low measurement accuracy and is difficult to apply to sitting posture analysis in automobiles. Aiming at the existing measurement problems, this paper proposes an improved three wavelength phase shift profilometry (TWPSP) method for measuring human body 3D data^[10-12]. The method combines the structured light with the marked points, and can quickly measure the positional angle data of the human body located in the automobile by positioning marks attached to the human body. According to this improved method, three kinds of fringe pattern with proper frequencies is projected on the target object. The wrapped phase can be calculated directly by these frequencies, without calculating the equivalent frequencies. It not only enhances the stability of the calculation process, but also avoid error propagation, and improves measurement accuracy. In addition, the method has good robustness, can achieve better measurement results in complex environments, and can overcome the weak reflectivity of black objects in laser scanning.

The remainder of this paper is organized as follows. Improved three frequencies phase shift profilometry method are explained in the Sec. 1. In Sec. 2, the preprocessing method of the human point cloud data, pasting position of the marking points, and the explanation of key angles are introduced. Sec.3 provides experimental results and analyses, and Sec. 4 concludes the paper.

1 Improved three frequencies phase shift profilometry method

The advantage of the phase shift method is that it can realize fast three-dimensional measurement, wide application scenarios, and high robustness. Its disadvantage is that it can not obtain better measurement results for metallic objects with dark ambient light or reflective properties. The human body pose data collected in this paper does not exist in the above two cases, so the phase shift method is suitable. The main principle of 3D reconstruction based on phase-shifted structured light is to project fringe images of different frequencies onto the surface of the target object. This kind of structured light is different from the normal gray code mode, and it has a certain sinusoidal property, which can make up for the defects of the gray code phase resolution mode. For the traditional three-step phase shift method^[13], there is a large number of arctangent operations in the phase expansion process, which results in higher phase expansion costs. The four-step phase shift method^[14]can

not effectively suppress the influence of uniform harmonics, and it is not sensitive to background brightness. This method can accelerate the reconstruction and improve the reconstruction accuracy. In order to obtain high-precision 3D data to meet the subsequent data processing, this paper adopts the six-step phase shift method, which can suppress the influence of the nonlinear response of the camera, thereby is helpful for the measurement accuracy.

Assuming that modulated images acquired by the camera are denoted by I_k . The sinusoidal fringe light with *i* frequencies and $\phi_i(x, y)(i = 1, 2, 3)$ is wrapped phase, the six-step phase-shift method adopted in this paper can be described as

$$I_{k}(x,y) = A(x,y) + B(x,y)\cos\left(\phi_{i}(x,y) + \frac{\pi \times k}{3}\right),$$

$$k = 0, 1...5$$
(1)

where A(x, y) is the background light intensity, B(x, y) is the modulation light intensity, and $\phi_i(x, y)$ can be calculated as

$$\phi_i(x,y) = \arctan\left(\frac{I_5(x,y) - I_3(x,y)}{(I_1(x,y) + I_4(x,y)) - (I_3(x,y) + I_5(x,y))}\right),$$

$$i = 1, 2, 3$$
(2)

where $\phi_i(x, y)$ belongs to $[-\pi, +\pi]$ and has a discontinuity of 2π .

For the existing three frequencies phase shift profilometry method^[15], it uses periods length of $\lambda_1 = 21$ pixel, $\lambda_2 = 18$ pixel and $\lambda_3 = 16$ pixel as the modulation period of the projected fringes. Based on these three basic periods, longer equivalent periods $\lambda_{12} = 126$ pixel, $\lambda_{23} = 144$ pixel, $\lambda_{123} = 1008$ pixel can be calculated. Equivalent periods can be calculated as

$$\lambda_{12} = \left| \frac{\lambda_1 \times \lambda_2}{\lambda_1 - \lambda_2} \right|, \ \lambda_{23} = \left| \frac{\lambda_2 \times \lambda_3}{\lambda_2 - \lambda_3} \right|, \ \lambda_{123} = \left| \frac{\lambda_{12} \times \lambda_{23}}{\lambda_{12} - \lambda_{23}} \right|$$
(3)

It should be noted that the final equivalent period needs to meet the conditions that its period is greater than the transverse resolution of the projector so that discontinuous conditions can be avoided. At the same time, the wrapped phase $\phi_{12}(i, j)$, $\phi_{23}(i, j)$ and $\phi_{123}(i, j)$ corresponding to the equivalent period can be calculated. Then, the corresponding unwrapped phase $\Phi_3(i, j)$ corresponding to $\lambda_3 = 16$ can be calculated as

$$\Phi_{3}(x,y) = \phi_{3}(x,y) + 2\pi \left(\text{INT}\left(\frac{\phi_{123}(x,y)}{2\pi} \times \frac{\lambda_{123}}{\lambda_{23}}\right) \times \frac{\lambda_{23}}{\lambda_{3}} + \text{INT}\left(\frac{\phi_{23}(x,y)}{2\pi} \times \frac{\lambda_{23}}{\lambda_{3}}\right) \right)$$
(4)

It can be seen from the above derivation process that there is a cumulative error in the process of calculating wrapped phases of different periods, which ultimately leads to the absence of the generated 3D data. In order to avoid the above situation, we select $\lambda'_1 = 1008$ pixel, $\lambda'_2 = 144$ pixel and $\lambda'_3 = 16$ pixel as the modulation periods of projected stripes and the corresponding wrapped phase $\phi'_i(x, y)$ can be calculated by Eq.(2), where $\phi'_1(x, y) = \phi_{123}(x, y)$, $\phi'_2(x, y) = \phi_{23}(x, y)$ and $\phi'_3(x, y) = \phi_3(x, y)$. In this way, the unwrapped phase $\Phi'_3(x, y)$ can be obtained without calculating the equivalent period. Usually, the phase can be calculated as

$$\theta_G(x, y) = \theta_w(x, y) + 2\pi * m(x, y)$$
(5)

where $\theta_G(x, y)$ is absolute unwrapped phase, $\theta_w(x, y)$ is wrapped phase, m(x, y) is an integer.

For $\lambda'_1 = 1\,008$ pixel periods, there exists λ'_1/λ'_2 discontinuities. As for $\lambda'_2 = 144$ pixel periods, there exists λ'_2/λ'_3 discontinuities. In order to eliminate the discontinuities, the unwrapped phase $\Phi'_3(x, y)$ is revised to

$$\mathfrak{P}_{3}^{\prime}(x,y) = \begin{cases}
\phi_{3}^{\prime}(x,y) + 2\pi \left(\mathrm{INT} \left(\frac{\phi_{1}^{\prime}(x,y)}{2\pi} \times \frac{\lambda_{1}^{\prime}}{\lambda_{2}^{\prime}} \right) \times \frac{\lambda_{2}^{\prime}}{\lambda_{3}^{\prime}} + \mathrm{INT} \left(\frac{\phi_{2}^{\prime}(x,y)}{2\pi} \times \frac{\lambda_{2}^{\prime}}{\lambda_{3}^{\prime}} \right) \right) \\
\phi_{1}^{\prime}(x,y) \neq 2\pi \text{ and } \phi_{2}^{\prime}(x,y) \neq 2\pi \text{ and } \phi_{3}^{\prime}(x,y) \neq 2\pi \\
\phi_{3}^{\prime}(x,y) + 2\pi \left(\mathrm{INT} \left(\frac{\phi_{1}^{\prime}(x,y)}{2\pi} \times \frac{\lambda_{1}^{\prime}}{\lambda_{2}^{\prime}} \right) \times \frac{\lambda_{2}^{\prime}}{\lambda_{3}^{\prime}} + \mathrm{INT} \left(\frac{\phi_{2}^{\prime}(x,y)}{2\pi} \times \frac{\lambda_{2}^{\prime}}{\lambda_{3}^{\prime}} \right) - 1 \right) \\
\phi_{1}^{\prime}(x,y) \neq 2\pi \text{ and } (\phi_{2}^{\prime}(x,y) = 2\pi \text{ or } \phi_{3}^{\prime}(x,y) = 2\pi) \\
\phi_{3}^{\prime}(x,y) + 2\pi \left(\mathrm{INT} \left(\frac{\phi_{1}^{\prime}(x,y)}{2\pi} \times \frac{\lambda_{1}^{\prime}}{\lambda_{2}^{\prime}} - 1 \right) \times \frac{\lambda_{2}^{\prime}}{\lambda_{3}^{\prime}} + \mathrm{INT} \left(\frac{\phi_{2}^{\prime}(x,y)}{2\pi} \times \frac{\lambda_{2}^{\prime}}{\lambda_{3}^{\prime}} \right) \right) \\
\phi_{1}^{\prime}(x,y) = 2\pi \text{ and } \phi_{2}^{\prime}(x,y) \neq 2\pi \text{ and } \phi_{3}^{\prime}(x,y) \neq 2\pi
\end{cases} \tag{6}$$

The corresponding phase order is calculated from the corresponding relationship of $\phi'_1(x,y)$, $\phi'_2(x,y)$ and $\phi'_3(x,y)$. When $\phi'_2(x,y)$ or $\phi'_3(x,y)$ generates phase jump and $\phi'_1(x,y)$ not generates, we subtract 2π from the unwrapped phase to compensate the error. When $\phi'_2(x,y)$ and $\phi'_3(x,y)$ have no phase jump and $\phi'_1(x,y)$ generates phase jump, we subtract $2\pi\lambda'_2/\lambda'_3$ from the unwrapped phase to compensate the error. When the phase jump is not generated, no compensation is required, as shown in Eq.(6).

The improved TWPSP method without calculating the synthesized phase, avoids the error generated during the phase combination, thereby reducing the error in understanding the parcel process, accelerating the process of understanding the phase, and improving the quality of the 3D point cloud data. Compared with traditional method, this method does not need to calculate the synthesized phase, and avoids error propagation in the calculation process. It can also effectively reduce the time of calculating the unwrapping phase and increase the speed of the entire 3D reconstruction process. It will be demonstrated that the method meets the measurement needs of this paper.

2 Data processing

Since the human body point cloud data acquired by the 3D scanning is contaminated by environmental noise, preprocessing is required after the measurement, in order to reduce the error and improve the measurement accuracy. The steps of preprocessing involved in this paper include denoising, coordinate system correction, and data analysis.

Before measurement, we need to paste some markers on the key positions of the human body, so that we can quickly extract and find the key coordinate points during data processing. The positions of the markers are shown in the Fig. 1(a).

Then we use the scanner to get human body point cloud data. But the scanning point cloud usually contains several kinds of defects or external noise. Therefore, it is necessary to preprocess the data. In the preprocessing stage, the octree^[16] is a commonly used data structure.





Fig.1 Target data (a) distribution of the markers (b) angle calculation diagram

which is used to obtain the neighborhood information of a point. The octree can improve search efficiency, and save processing time as well. This paper uses octree to traverse the point cloud data obtained through scanning and delete invalid point cloud data.

For each point, calculating its average distance to a specified number of adjacent points. If the average distance is outside the standard range, it can be classified as an outlier and removed from the data. The average distance can be calculated as

$$D = \left(\sum_{i=1}^{k} \sqrt{(x - x_i)^2 + (y - y_i)^2 + (z - z_i)^2}\right) \frac{1}{k}$$
(7)

where k is the number of adjacent points, and x, y and z denote the coordinate of the current point of target point cloud.

After performing point cloud filtering, it is also necessary to transform all point cloud data which is collected in different scenes into a unified coordinate system. Rodriguez's rotation equation is to calculate the rotation matrix between two vectors from the angle and the rotation $axis^{[17]}$. This paper mainly uses the main element analysis method^[18] to calculate the vector *v* of the main direction of the target point cloud. The angle between the vector *v* and the main direction vector *u* of the standard coordinate system can be calculated as

$$\theta = a \cos\left(\frac{v_1 u_1 + v_2 u_2 + v_3 u_3}{\sqrt{v_1 v_1 + v_2 v_2 + v_3 v_3} + \sqrt{u_1 u_1 + u_2 u_2 + u_3 u_3}}\right)$$
(8)

Where θ is rotation angle, v is the main direction vector of the original point cloud, and u is the main direction vector of the standard coordinate system.

The revolution axis vector can be calculated as

$$\binom{r_1}{r_2} = \binom{v_2 u_3 - v_3 u_2}{v_3 u_1 - v_1 u_3}_{v_1 u_2 - v_2 u_1}$$
(9)

According to the rotation axis and the rotation angle,

the rotation matrix *R* can be calculated as

$$R = \begin{bmatrix} r_1^2 + (1 - r_1^2)\cos\theta & r_1r_2(1 - \cos\theta) + r_3\sin\theta & r_1r_3(1 - \cos\theta) + r_2\sin\theta \\ r_1r_2(1 - \cos\theta) - r_3\sin\theta & r_2^2 + (1 - r_2^2)\cos\theta & r_2r_3(1 - \cos\theta) - r_1\sin\theta \\ r_1r_3(1 - \cos\theta) - r_2\sin\theta & r_2r_3(1 - \cos\theta) + r_1\sin\theta & r_3^2 + (1 - r_3^2)\cos\theta \end{bmatrix}$$
(10)

where r is revolution axis vector.

In order to translate the target point cloud to the origin position of the standard coordinate system, it is necessary to calculate the center of gravity of the target point cloud. Point of gravity can be calculated as

$$\binom{C_x}{\binom{C_y}{C_z}} = \begin{pmatrix} (\sum_{i=1}^n x_i)/n \\ (\sum_{i=1}^n y_i)/n \\ (\sum_{i=1}^n z_i)/n \end{pmatrix}$$
(11)

where *n* is the number of target point cloud, x_i , y_i and z_i is the point of target point cloud.

The translation matrix $T(T_x, T_y, T_z)$ can be calculated as

$$\begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix} = \begin{bmatrix} -c_x \\ -c_y \\ -c_z \end{bmatrix}$$
(12)

After calculating the rotation matrix and the translation matrix, we can rotate and translate the target point cloud data to the standard coordinate system. The corrected coordinates of point cloud P' can be

transformed by

$$\begin{bmatrix} P'\\1 \end{bmatrix} = \begin{bmatrix} R & T\\0^T & 1 \end{bmatrix} \begin{bmatrix} P\\1 \end{bmatrix}$$
(13)

where P is the number of target point cloud.

After completing the above steps, the next step is to calculate the angle among key points and complete the data analysis. The angle calculation diagram of the 3D human body inside car is shown in Fig. 1(b).

In order to complete the calculation of the key angle, distance between two adjacent key points can be calculated as

$$d(A,B) = \sqrt{(x_A - x_B)^2 + (y_A - y_B)^2 + (z_A - z_B)^2}$$
(14)

where $A(x_A, y_A, z_A)$, $B(x_B, y_B, z_B)$ represent two key points of target point cloud.

After getting the distance between the key points, the next step is to calculate the space vector of the adjacent key points. Space vector can be calculated as

$$\overline{AB} = (B_x - A_x, B_y - A_y, B_z - A_z)$$

$$\overline{AC} = (C_x - A_x, C_y - A_y, C_z - A_z)$$
(15)

where $A(x_A, y_A, z_A)$, $B(x_B, y_B, z_B)$, $C(x_C, y_C, z_C)$ represent key points of target point cloud.

$$\cos\theta_s = \frac{AB \cdot AC}{|AB||AC|} \tag{16}$$

where θ_s is the key angle value of the human sitting posture.

3 Experimental results and analysis

A 3D measurement and reconstruction system for human body inside car is developed in our lab, including both software and hardware. The binocular 3D human body scanning system is shown in Fig. 2, where a projector with resolution of 1 024×768 is used to cast three kinds of sine stripes and the three sine photo periods are 1 008 pixel, 144 pixel, and 16 pixel, respectively. The industrial camera sends a 1 280×1 024 gray image to the computer at a frame rate of 4 frames/s. The industrial camera uses an 8 mm lens. As shown in Fig. 2, a person sits in a car and adjusts his seat to the most comfortable posture when driving, scans the human body in this posture to obtain the three-dimensional posture information of the human body. In order to ensure the integrity of the data, the scanning range is $1.5 \text{ m} \times 2 \text{ m}$, and the binocular 3D measurement system is more than 2 m away from the human body.



Fig.2 Binocular 3D human body scanning system

To examine the performance of the improved TWPSP method. In the same distance and scanning ranges we use the traditional TWPSP method^[15] and the method proposed in this paper to project the three kinds of calibrated balls with diameter of 20 mm. These three calibration balls are made of matte materials, which can suppress light reflection. The accuracy of the calibration sphere can reach 0.000 1 mm.

In Fig.3(a), the calibration ball are placed in different areas for 3D scanning to obtain its corresponding 3D reconstruction data in different areas. In Fig. 3(b), the distance between the centers of the three balls is 25 mm, 50 mm and 55.90 mm. We use the measurement scheme shown in Fig. 3(a) to obtain the measurement accuracy of 3D scanning system and fit the calibration ball 3D point cloud data obtained by two methods.

Through the measured 3D data, calibration Ball 1, calibration Ball 2 and calibration Ball 3 can be fitted respectively by Geomagic studio, and obtain the diameters \hat{d}_1 , \hat{d}_2 and \hat{d}_3 of the three calibration balls, and the sphere centers \hat{O}_1 , \hat{O}_2 and \hat{O}_3 of the three calibration balls.

In the next moment, the distance between the sphere



Fig.3 Calibration balls (a) calibration balls used in system measurement accuracy (b) schematic diagram of the spacing of the calibration balls

centers of three calibration balls $\hat{O}_1\hat{O}_2$, $\hat{O}_2\hat{O}_3$ and $\hat{O}_1\hat{O}_3$ can be obtained by the coordinates of \hat{O}_1 , \hat{O}_2 and \hat{O}_3 . We made ten measurements and calculated average distances and RMSE of $\hat{O}_1\hat{O}_2$, $\hat{O}_2\hat{O}_3$, $\hat{O}_1\hat{O}_3$, and the first five measurements and calculations are shown in Tab. 1.

As can be seen from Tab. 1, the RMSE of the spherical distance of the three calibration balls measured by the traditional method is 0.044 mm, 0.053 mm and 0.055 mm. RMSE of the spherical distance of three calibration balls measured by the improved method proposed in this paper is 0.019 mm, 0.037 mm and 0.035 mm. By comparing the experimental results, it can be seen that the measurement results obtained by the improved method have higher stability.

To further validate the effectiveness of the improved method proposed in this paper, three standard balls were

Key dimensions		Traditional TWPSP method	Method in this paper	
	1 2	49.853 50.053	49.970 50.038	
	3	49.979	50.000	
$\hat{O}_1\hat{O}_2/mm$	4	49.956	50.009	
0102	5	50.014	50.024	
	Average	49.971	49.998	
	RMSE	0.044	0.019	
	1 2	25.096 24.975	25.072 25.019	
	3	25.048	24.963	
$\hat{O}_2\hat{O}_3$ /mm	4	24.981	25.051	
	5	2525.077	24.949	
	Average	25.061	25.028	
	RMSE	0.053	0.037	
	1 2	55.654 55.643	55.924 55.934	
	3	55.572	55.910	
$\hat{O}_1\hat{O}_3$ /mm	4	55.632	55.881	
	5	55.610	55.875	
	Average	55.940	55.895	
	RMSE	0.055	0.035	

Tab.1 First five measurements of calibration balls

used for ball calibration experiments. The experimental results are shown in Tab. 2. As can be seen from Tab. 2, the mean absolute error (MAE) of each ball is solved respectively and \hat{d}_1 , \hat{d}_2 and \hat{d}_3 is the average of ten measurements. The mean of the MAE of the three calibration balls is taken as the average measurement accuracy of the system. The average measurement accuracy of the system using the improved method is 0.021 mm. Compared with the measurement accuracy calculated by the traditional method, the improved method has higher measurement accuracy, so in the actual measurement, the improved method can be used to obtain higher precision data.

The parts within the red rectangles in Fig. 4(c) and Fig. 4(d) show the difference between the two methods. The 3D point cloud data obtained by the traditional method has a case of mismatching. In contrast, the point cloud data obtained through the improved method is more dense and the data is more accurate.

	d_1 /mm	d_2 /mm	\hat{d}_3 /mm	MAE_1 /mm	MAE_2 /mm	MAE ₃ /mm
Traditional TWPSP method	19.949	19.961	19.978	0.048	0.035	0.039
Method in this paper	19.979	19.978	19.988	0.028	0.018	0.019
(a) $I_k(i_3j)$ $\lambda_1=21$ $\lambda_2=18$ $\lambda_3=16$	$\Phi_{1}(i,j)$ $\Phi_{2}(i,j)$ $\Phi_{3}(i,j)$		(c)	France Acceptor and 2001	(c)	
(b) $\lambda_1 = 1\ 008$ $I_k(i,j)$ $I_k(i,j)$ $\Phi_1(i,j)$	$\lambda_2 = 144$	$\lambda_3 = 16$	(d)		(f)	

 Tab.2 Data comparison table of fitting results

Fig.4 Method comparison (a) traditional TWPSP body scanning schematic (b) improved TWPSP body scanning schematic (c) fitting result of traditional method (d) result of the method in this paper (e) traditional TWPSP method reconstruction chart (f) improved TWPSP method reconstruction chart

Next, the paper further compares the reconstruction effects of the two reconstruction methods in real anthropometric scenarios. The measurement distance is more than 2 m, and the measurement range is $1.5 \text{ m} \times 2 \text{ m}$. Different ways of rebuilding flowchart are shown in the Fig. 4(a) and Fig. 4(b).

Experiments show that point cloud data obtained by the improved phase unwrapping method is more compact, and the details of the point cloud data are more perfect, and no details are missing. The reconstruction effect of the two methods is compared as shown in Fig. 4(e) and Fig.4(f).

To further evaluate the performance of the scanning system using the method presented herein. In this paper, we use the commonly used body scanner to collect data in the case of the same person and vehicle. The Artec Eva 3D scanner does not require marking or calibrating objects. It can capture the shape of the object with high resolution and restore the bright colors of the object. The Sense scanner is a laser scanner that has a short scan time and is capable of quickly generating 3D information of the target. This paper collects human point cloud data using several different scanners and compares the number of generated point clouds, accuracy, and reconstruction time. The comparison results are shown in Tab.3.

As seen from Tab.3, the number of point clouds generated by Artec scanners is the highest, but Artec's accuracy is only 0.1 mm, and its reconstruction time takes 7 s. Although the reconstruction time of the sense sensor is 2 s lower than Artec, it generates sparse point clouds with an accuracy of only 0.9 mm, which is insufficient for measurement needs. The point cloud data collected by the method proposed in this paper not only has an accuracy of 0.03 mm, but also requires only 1 s for reconstruction

Tab.3 Point cloud data analysis of different sensors							
Equipment	Point cloud number	Point cloud density	Accuracy/mm	Reconstruction time/s			
Artec scanner	795 785	0.143	0.1	7			
Sense scanner	121 717	0.001	0.9	5			
Traditional TWPSP method system	236 540	0.016	0.055	2			
Scanning system of this paper	296 540	0.016	0.03	1			

time. Through the above comparison, it can be concluded that the improved TWPSP method proposed in this paper has a fast reconstruction speed and a high point cloud density, which is suitable for fast 3D anthropometry. It can be demonstrated that the improved optical profilerbased optical 3D body scanner has better accuracy and point cloud quality.

We used the method proposed in this paper to collect 20 sets of human body 3D data on three different models of car. For each set of data for each car, the average of 9 key angles were calculated. We also calculated the average of the angles of each type and RMSE. The calculation results are shown in Tab.4. It can be seen from the Fig.5 that the angle pose data of the 20 experimental objects on different cars are relatively close, and the fluctuation range is small. The largest RMSE of the 3D measurement method for human body inside car proposed in this paper in different cars is 8.661 mm.

	Average degree	Average degree	Average degree	RMSE of	RMSE of	RMSE of
	0110.1 cai	01110.2 car	01110.5 cai	NO.1 Cal	110.2 Cal	110.5 cai
Angle A1	25.102	26.410	23.605	1.291	1.454	1.249
Angle A2	98.491	104.70 5	98.370	5.230	5.775	5.081
Angle A3	60.806	64.313	65.654	3.426	3.804	3.117
Angle A4	101.501	93.416	101.328	5.966	5.320	5.313
Angle A5	35.543	30.677	35.385	3.926	3.165	3.381
Angle A6	127.434	133.667	138.095	2.117	2.084	2.879
Angle A7	154.080	141.380	144.400	8.344	8.661	8.250
Angle A8	41.647	32.875	43.192	6.775	6.625	6.291
Angle A9	13.935	9.661	14.399	3.864	3.297	3.958

Tab.4	Angles of key	points (all	values are	reported as	degree)
1	This of hey	pomes (an	ranaes are	reported as	uchicc)



Fig.5 Comparison of the average values of the three groups of poses

As shown in Tab. 4, the maximum value of RMSE in the measurement results is 8.661 mm, and the minimum value of RMSE is 1.249 mm. The experimental results show that the improved TWPSP method can realize fast human body 3D measurement to obtain more complete point cloud data. In addition, the key point angle data calculated by the points measured by this method is highly robust.

Conclusion 4

This paper proposed a fast 3D measurement method for 3D measurement of human posture. Compared with traditional three-frequencies 3D measurement method, the improved method proposed in this paper has higher measurement accuracy and reconstruction speed. When 第6期

the distance is more than 2 m and the measuring range is $1.5 \text{ m} \times 2 \text{ m}$, the reconstruction accuracy can reach 0.03 mm, and the reconstruction time only takes 1 s. In addition, compared with the existing handheld 3D scanning equipment, the 3D scanning system constructed by the improved method has higher point cloud quality, faster reconstruction speed and moderate point cloud density. At the same time, the key angle data obtained by this method has small fluctuations and high stability of repeated measurements, which meets the needs of 3D measurement of human posture. Our future work includes automatic and rapid extraction of human body landmarks, combining 3D measurements with artificial neural networks.

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