

Research and Application of Improved Hough-ICP Algorithm

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Abstract: During the vehicle identification in transportation, many problems must be overcome: firstly, modern vehicle outline leads the failure of getting features from 3D rigid body. Secondly, synthesis distortion is hard to overcome in contour connecting because of front wheels torsion. So a new algorithm of 3D synthesis is presented by improved Hough and ICP analysis to solve these problems. In the stage of getting three-view drawing, Hough analysis is used to transform circle signals to line signals which are the component of newly vehicle outline, at the same time, oval signals of front wheel are fitted based on back wheel. In the contour connecting stage, a new concept named projection wall is proposed during the construction of covariance matrix based on ICP analysis, and then, the maximum values of projection walls are calculated by four-element method analysis. Finally, 3D rigid body is described accurately. Experimental results show that the proposed algorithm not only reduces the noise cased by vehicle accessories, but also overcomes signal distortion and time serious.

Introduction

At present,3D synthesis belongs to high dimensional matrix category, the popular vehicle 3D research focus on the application in the vehicle body design and auto component parts processing, including size/fluid detection, motion and strain analysis, component visual simulation^[1], but few studies have been done on the classification of vehicles. Reference [2] uses the perspective distortion modeling, and the 3D information obtained from 2D information is classified by SVM classifier; reference [3] using anisotropic magnetoresistive sensors and other auxiliary equipment, vehicles are classified by BP neural network. The rest of the "category" detection methods include electromagnetic wave detection ^[4], fuzzy neural network ^[5]. These methods are based on multi sensor proofread 3D environment simulation, and the default function signal must be in line with the prior knowledge, which do not conform to the type of vehicle check in Guangzhou city, because Guangzhou has no prior knowledge of vehicle security, just according to the detection of the vehicle model and 3D information to verify the vehicle model and cargo capacity.

Therefore, this paper presents an improved Hough-ICP 3D synthesis algorithm based on the types of vehicle sign, which analysing outline signal through the Hough, and fitting the front wheel to get a modified three view signal, at the same time, the use of ICP analysis of the "projection wall" to achieve 3D contours seamlessly, and finally realize the detection of the vehicle type.

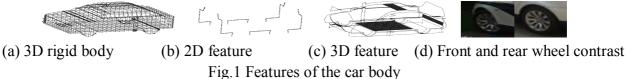
Car body feature description

Car body description

Here the paper defines: the feature that in each orthographic projection of the "three views"

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(the main view, left view, and top view) have intersection is called 3D rigid body features, and the feature that in two orthographic engineering drawing of the "three views" have intersection is called 2D rigid body features. As shown in Fig.1, the car 3D rigid body features mainly distributed in the edge of the chassis, front and rear bumper side edges; compared with the features of 3D rigid, 2D rigid features would be more abundant, which mainly distributed in the chassis, the roof and the front and rear bumper. The study found that 2D and 3D rigid body features can indicate the type of car only, therefore, how to overcome the weakness of the 3D rigid body features caused by the streamline shape of the modern automobile, and how to realize the seamless connection of the 3D profile of the vehicle model, is the key to get the features of the 3D rigid body. At the same time, the front wheel twist will make the 3D synthesis distortion (as shown in Fig.1 (d)) which is difficult to overcome.



Conception of projection wall

As shown in Fig.2, the independent three view signal, because of the difference of the angle, distance, the gap and the coordinate system which is established according to the acquisition system, it leads to the mismatch between the two sides of the edge signal, or can not reach the threshold of matching, therefore, it needs to seek the maximum solution in 3D space. This paper defines: the three view signal along the X, Y, Z vertical extension, the aggregate of 3D signals which is surrounded by four sides is called "projection wall".

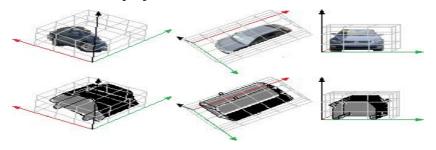


Fig.2 Concept of projection wall

Improved Hough-ICP algorithm

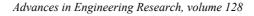
Fitting correction

Definition 1: suppose there is a set F in the X-Y image space, there is a collection of P in the A-B-R parameter space. Define the set of integers $S = \{s_i\}, i \in R$, satisfies the formula (1), and

 $s_i \neq 0$. The Sum(·) is a statistic which satisfies the equation.

$$s_i = Sum((x-a)^2 + (y-b)^2 = r^2)$$
(1)

Definition 2: suppose there is a set F in the X-Y image space, there is a collection of P in the A-B parameter space. Define the set of integers $S = \{s_i\}, i \in R$, satisfies the formula (2), and





 $s_i \neq 0$. The Sum(·) is a statistic which satisfies the equation.

$$s_i = Sum(ax + b = y) \tag{2}$$

Feature extraction of projection wall Characteristic vector

Definition 3: suppose F_0 , F_1 , F_2 are respectively the left view, top view, the main view of the 2D vincent^[6] coefficient matrix, and meet the conditions of 1. Its characteristic vector is defined as

$$V = [V_0, V_1, V_2]^T$$

$$V_i = [G(F_i)]^T$$
(3)

And G(.) is the 3D Vincent coefficient matrix which is extended along the direction perpendicular to the 2D plane of F_i .

Condition 1: suppose there is a set of integers $F = \{f_1, \dots, f_i\}, i \in \mathbb{R}$, lets the formula

$$\sum_{m=-1}^{m=1} \sum_{n=-1}^{n=1} f(x+m, y+n) \le 3 \text{ set up. While } f(x, y) \text{ is the pixel value of the coordinate } (x, y).$$

ICP feature points

Definition 4: suppose there are set P and X, and the set satisfies $P \cap X = N \neq \phi$. Define coordinate transformation matrix:

$$q = [q_R | q_T]$$
(5)
min $f(q) = \frac{1}{N} \sum_{i=0}^{N} ||x_i - R(q_R)p_i - q_T||^2$ (6)

Where q_R is 3 × 3 rotation matrix, q_T is the unit column vector translation matrix,

min $f(q) \neq \phi$ is the minimum value of the transformation matrix q.

Design of algorithm flow

Step 1: vincent edge extraction of the motion blurred image signal, and eliminates non single pixel. According to the three views, the corresponding 3D coordinate system is set up, and the 3D plane signal is generated in the X - Y - Z image space.

Step 2: according to formula (1) to solve the parameter space of $(a_{max}, b_{max}, r_{max})$, and the

corresponding coordinates F_i of the image space signal. Let it is the standard, rectify the front

wheel which swing into the ellipse to circle. And according to the formula (2), using the same method to fit a straight line of outline signal.

Step 3: according to the formula (3) and (4) to construct the feature vector of the projection

wall, according to the formula (7) to construct the covariance matrix of F_0 , F_1 .



$$\sum_{F_0,F_1} = \frac{1}{X} \sum_{i=0}^{N} \left(o_{F_0} - \mu_{F_0} \right) \left(o_{F_1} - \mu_{F_1} \right)^T$$
(7)

where $o = \{x, y, z\}$, $\mu = \{\mu_x, \mu_y, \mu_z\}$, $\mu_x = m_{100}/m_{000}$, $\mu_y = m_{010}/m_{000}$, $\mu_z = m_{001}/m_{000}$, m_{xyz} can be

obtained by formula (8).

$$m_{xyz} = \frac{1}{X^{x}Y^{y}Z^{z}} \sum_{i=0}^{X} \sum_{j=0}^{Y} \sum_{k=0}^{Z} i^{x} \times j^{y} \times k^{z} \times G(i, j, k)$$
(8)

Step 4: according to (5), the transformation matrix is constructed, and the characteristic value is calculated. The characteristic vector corresponding to the maximum characteristic value is the best rotation vector which is $q_R = [q_0 \quad q_1 \quad q_2 \quad q_3]^T$, and the best translation vector is

 $q_T = \mu_{F_0} - R(q_R)\mu_{F_1} \cdot R(q_R)$ can be obtained by formula (9).

$$R(q_R) = \begin{bmatrix} q_0^2 + q_1^2 - q_2^2 - q_3^2 & 2(q_1q_2 - q_0q_3) & 2(q_1q_3 + q_0q_2) \\ 2(q_1q_2 + q_0q_3) & q_0^2 - q_1^2 + q_2^2 - q_3^2 & 2(q_2q_3 - q_0q_1) \\ 2(q_1q_3 - q_0q_2) & 2(q_2q_3 + q_0q_1) & q_0^2 - q_1^2 - q_2^2 + q_3^2 \end{bmatrix}$$
(9)

Adjust the F₀ and F₁ according to q_R and q_T to make it possible to coincide with one side of the projection wall. And according to the formula (6) to obtain the minimum value. if min f(q) < 0.001, then F₀ and the overlap of one side of the projection wall as a benchmark, adjust the proportion F₁, so that the "opposite" of the projection wall also coincide.

The same method is used to deal with the F_3 , which makes the projection wall of the four sides coincide with the projection wall composed of F_0 and F_1 , to get car body signal.

Step 5: 3D edge detection of the car body signal to extract the 3D contour signal. 2D and 3D rigid body features $E = \{E_0, E_1\}$ are separated by calculating the number of overlapping points in the 3D contour signal. At the same time, by using 3D moment invariant ^[7], weighting the $\{E_0, E_1\} = \{0.15, 0.85\}$, and analysis the length, width and height of the vehicle body obtained by the acquisition equipment, finally get the motorcycle type.

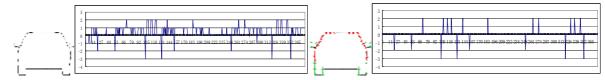
Test and analysis

Hough fitting analysis

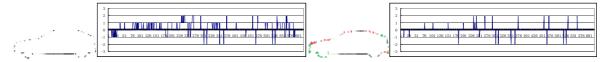
Experiment 1 is to verify that the algorithm is robust to overcome the streamline signal. Fig. 3 (a), (c), (E) are three views of the original signal and curvature analysis, it can be seen that the curvature is not "0" is signal points with "curve", the sharp signal which number is "2" and "3" is the position that existence of edges and corners or curve trend changed significantly. Fig.3 (b), (d), (f) are the three views Hough fitting signal and curvature analysis, the contour signals of three visual icon " \times " and " Δ " are linear fitting and circle fitting part based on the algorithm. It can be seen from the curvature of the corresponding analysis that in addition to the contour edges, other noise "curve" signal is eliminated, and replaced by "linear signal", so as to maximize the



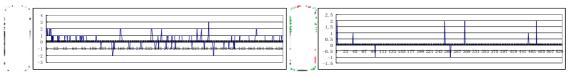
elimination of 3D rigid body characteristics caused by the lack of streamlined shape, so that it makes it possible to analyze the "overlapping" of the "projection wall" on the following steps.



(a) Analysis of the original signal and curvature of the main view (b) Analysis of Hough fitting signal and curvature of the main view



(c) Analysis of the original signal and curvature of the left view (d) Analysis of Hough fitting signal and curvature of the left view



(e) Analysis of the original signal and curvature of the top view (f) Analysis of Hough fitting signal and curvature of the top view

Fig.3 Hough fitting analysis of three views

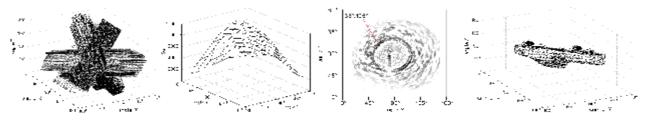
Projection wall ICP synthesis

Experiment 2 aims to present the process of ICP synthesis of the projection wall and verify its effectiveness. Fig. 4 (a) is for the three plane projection wall in a unified coordinate system according to the synthesis of the initial coordinates, the extension signal G_m roate around the axis of X, Y and Z respectively and adjust according to proportion in 3D space, any two projection walls in each rotation vector q_R has a maximum overlap area. Figure 4 (b) shows overlapping pixel

statistics which is get by one of the projection wall rotating around the axis of X, Y and adjusting scaling factor, it can be seen that on the same scaling factor, a pair of projection wall in a plurality of rotation angle will form a coincident extremum, so as to form a mutual connection, similar to "the back" of mountain ridge, which will have a maximum value. Fig.4 (c) is the projection drawing

of Fig. 4 (b) in X, Y planes, it is intuitive to see that the maximum value is $(63^{\circ}, 105^{\circ})$. Fig.4 (d) is

the final composite image obtained by ICP analysis of the projection wall.



(a) Specification projection wall coordinates
 (b) ICP synthesis of extreme value statistics
 (c) ICP synthetic extreme projection
 (d) Final composite image
 Fig.4 Synthesis analysis of projection wall ICP

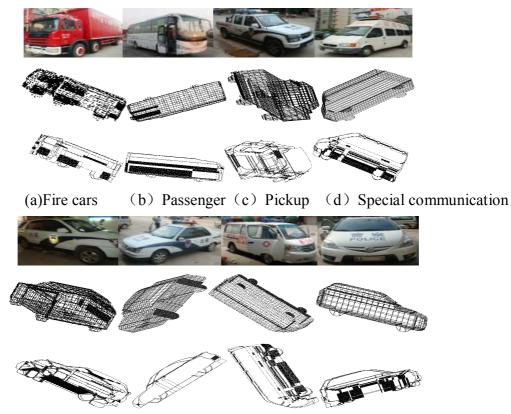


Hough-ICP analysis

Experiment 3 aims to verify the universality of the algorithm. Relying on the support of the Guangzhou 110 emergency center, this study obtained 17 fire cars, 55 medical cars, 65 police cars, a total of 137 experimental samples covering the fire cars, passenger , pickup ,special communication, cross-country, three car, ambulance and hatchback. The proposed algorithm is used for 3D synthesis and category detection. The 3D contour map of the second row of figure 5 (a) \sim (H) shows that the rigid characteristics of vehicle accurately sketched out, streamlined body characteristics are ignored, and the noise signals which influence the detection effect, such as alarm lamps, rearview mirrors, receiving antenna board of special communication are eliminated. At the same time, the third row of figure 5 (a) \sim (H) shows that 3D rigid features (black entity graph area in figure) distribute chassis, front and rear bumpers and side edges, while the 2D rigid features (light dashed lines) are lie scattered in vehicle body, this is the reason of weighting 2D and 3D

signal E_0, E_1 . According to the weighted analysis of E_0, E_1 , it can lay a solid foundation for

vehicle classification detection.



(e) Cross-country (f) Three car (g) Ambulance (h) Hatchback Fig.5 Vehicle detection and signals of E_{0} , E_{1}

Table 1 shows the test datas of various cars, the 3D moment invariants obtained by the proposed algorithm. The length, width, highth datas obtained by acquisition equipment parameters conversion, these 4 kinds of datas are the average of 137 samples and covers the maximum and minimum value, from this comprehensive analysis can match the corresponding model. It can be seen from the table that the difference of the sample of the fire engine is bigger, so its parameter is floating big, the other car models are stable in the small interval.



Parameters Car types	3D moment invariants	length (m)	width (m)	highth (m)
Fire cars	23.54782±0.24286	9.73±2.33	3.21±0.53	3.26±0.42
Passenger	21.02467±0.01473	9.22±0.73	3.10±0.23	3.10±0.31
Pickup	9.12687±0.01224	4.11±0.30	1.30±0.21	1.30±0.20
Special communication	12.95186±0.01695	4.20±0.24	1.44±0.21	1.92±0.20
Cross-country	11.34568±0.01138	4.12±0.31	1.54±0.12	1.31±0.22
Three car	10.71439±0.01436	4.33±0.10	1.42±0.20	1.21±0.21
Ambulance	13.75169±0.02130	4.20±0.21	1.42±0.31	1.69±0.20
Hatchback	10.11245±0.01716	4.20±0.10	1.41±0.20	1.21±0.21

Table1 Vehicle inspection dates

Conclusion

This paper has the characteristics of foresight and exploration, the proposed improved Hough-ICP algorithm for 3D synthesis can detect vehicle type. Aim at 3D rigid body features missing and contour synthesis distortion which caused by streamlined body and the twist of the front wheel, the algorithm through Hough analysis outline signals with streamlined feature, and use the after wheel as the basic circle to fitting front wheel; using the ICP analysis "projection wall" to find the extremum to achieve 3D contour profile seamlessly, finally achieve the purpose of detecting vehicle type. It is worth mentioning that the design of the algorithm without any other auxiliary software and special hardware, but the speed up of the algorithm depends on the robustness of the bearing system.

Acknowledgements

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