NEURAL NETWORK BASED MULTISENSOR FUSION IN A NOVEL PERMANENT MAGNET MULTI-DOF ACTUATOR ORIENTATION DETECTION SYSTEM

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Abstract- The methods of multi-DOF actuator orientation detection based on machine vision and magnetic field sensor and neural network based multisensor fusion are presented in this paper. Special grid pattern is printed on the surface of rotor according to the pseudo-random encoder for the camera capturing the image of the rotor. The magnetic field sensors are adopted and properly placed circling the shaft of rotor for detection the orientation information. By analyzing the image of the rotor, the coordinates of characteristic points in the coordinate system fixed on the rotor are derived. Combined with the magnetic sensor scheme to reduce the measurement error and enhance the fault tolerant ability, a multisensor data fusion scheme using BP neural network is developed and validated by experiment. The results show that multisensor data fusion based on NN is superior to single
measurement method with more accuracy and high reliability, which is more effective and practical for applications.

Index terms: Permanent magnet, multi-DOF, orientation detection, machine vision, magnetic sensor.

I. INTRODUCTION

With the development of the industry and technology, robot and manipulators which can achieve three-degree-of-freedom (3-DOF) motion are used more and more widely. This kind of devices are usually built with several convention driver motor, each having single degree-of-freedom (DOF), which reduces the position accuracy, efficient, dynamic performance of the system. In this condition, the multi-DOF actuators or motors have attracted many attentions. To realize precision control, the feedback control system is necessary. However, the measurement systems used for single-DOF feedback control system is not suitable for the multi-DOF feedback control system. So the orientation detection becomes a crucial problem and should be paid more attentions. There are mainly five methodologies to measure multi-DOF orientations recently. These measurement methods are classified into two kinds contact and noncontact methods. Because of the flaws of contract measurement method, using non-contract method is the development trends of the orientation measurement.

K.M. Lee et al developed a contact measurement system that used three single-axis encoders for measuring 3-DOF motions [1]. It required a mechanism which introduced friction and inertia to constrain the device. D. Stein used 96 point sensors to measure the orientation of the rotor based on the color detected by the sensors [2-4]. The rotor was painted in two colors. It was difficulty to transfer the pattern to the rotor and used too many sensors. K.M. Lee and D. Zhou presented a non-contract optical sensor for real-time orientation measurement [5]. The optical sensors detected the microscopic changes of surface feature. The surface property of the rotor influenced the resolution of the measurement system. W. Wang et al used six hall devices to detect the orientation of the rotor according to the magnetic field due to the rotor permanent magnets [6,7]. The measurement accuracy was affected by the magnetic field generated by the stator winding current, so the sensor outputs should be compensated according to the current. Using vision based sensor to detect the orientation was firstly proposed by H. Garner et al. [8]. Different from the
linear and the circular strings designed by H. Garner, Z. Qian et al. presented a new grid pattern design and construction method using pseudo-random-sequence (PRS) [9]. The size of the windows limited the measurement accuracy of these vision based method, because spherical curved plane is equivalent to flat plane in image processing procedure for extracting information of the characteristic points.

In this paper, an effective pattern design method and image matching algorithm to avoid the equivalence for improving the detection accuracy are presented combined with the 3-DOF magnetic field sensor based detection method. The two modes both have the reliability and outer disturbance problems, so neural network based multisensor fusion methods are adopted. The camera based machine vision measurement calibrated data are used as training samples for the magnetic sensors modeling and complete data fusion and the error analysis in fault cases is carried out by experiments.

II. OPERATION PRINCIPLE OF THE PM MULTI-DOF ACTUATOR

The motor prototype consists of a ball-shaped rotor with four layers of 40 poles and two layers of 24 poles. There are 10 poles with equally spaced position of alternative N-S distribution for the rotor. There are 12 poles in each layer and controlled separately for the stator. Figure 1 shows the basic structure of the actuator. The actuator is designed to implement maximum 67 degrees tilt motion and 360 degrees rotation. The illustration of the torque produced by one pair of poles can be shown in figure 2.

![Figure 1](image)

(a) Side view of the stator and rotor, (b) Matching arrangement of the stator and rotor poles
Using the Lagrange energy method, the dynamic equations of multi-DOF actuator are obtained. Each equation represents an axis of Cardian Angle. It can be seen that there are many one-order and two-order coupling terms in the equation, which indicates serious inter-axes nonlinear coupling of the spherical actuator.

![Diagram of the actuator](image)

**Figure 2.** The illustration of the torque produced by one pair of poles
(a) The stator/rotor pole pairs’ positions, (b) The interaction torque between one pole pair

III. ORIENTATION DETECTION METHODOLOGY AND MULTISENSOR FUSION SCHEME

a. Machine vision based image orientation method
To obtain the rotor position signal by image analysis, the first step is coating the rotor surface with a characteristic pattern. Stevenson and Jordan firstly introduced the flat grid to the position measuring system for the measurement of automatic navigation vehicle position offset [15], but this method can only measure the offset of the two directions and cannot measure the amount of rotation. In this work, with the unequal spacing of grid lines on the part of rotor surface with the spraying method, the position features of the rotor can be loaded into the image. Grid spacing to be ranked is in accordance with the principle of pseudo-random coding. Figure 3 shows the plane and spherical surface grid to generate pseudo-random graphs. Diagram of the grid does not use the Earth’s latitude and longitude interval distribution, which is to guarantee the uniqueness of the rotor in an arbitrary rotation angle and direction of image features.

The vision sensor selects a region of image for the rotor surface to eliminate the effects of spherical curvature deformation, and only part of the extraction of the central region image is analyzed and calculated. The selected area is smaller and can be approximated to a plane processing, thus avoiding the image deformation. The straight-line distance available for matching feature points can take instead of the spherical distance.

Like the vessel tracking method, the feature points are obtained from the vision sensor rotor image (intersection center point of grid lines). Extract two-phase vertical grid lines in the grid obtained within the central region as the search direction. After matching, according to the location of these points in the original sequence, the latitude and longitude in the rotating coordinate system can be determined and also the coordinates in this coordinate system.

![Figure 3. Schematic Grid pattern on the plane and rotor surface](image)

The vision sensor calibration is to establish the relationship between the sensor pixel location and scene locations. The sensor model parameters can be solved according to the known feature
points’ image coordinates and world coordinates. The initial position of the world coordinate system fixed at the center of the sphere of spherical rotor, even with the solid overlap in center of the rotor rotating reference frame. In this paper, the linear model of the sensor is used to simplify the computation to estimate the sensor parameters.

\[
\begin{bmatrix}
  u_i \\
  v_i \\
  1
\end{bmatrix} =
\begin{bmatrix}
  m_{11} & m_{12} & m_{13} & m_{14} & X_i \\
  m_{21} & m_{22} & m_{23} & m_{24} & Y_i \\
  m_{31} & m_{32} & m_{33} & m_{34} & Z_i
\end{bmatrix}
\]

(1)

where \( u_i, v_i \) are the pixel points in the image, \( X_i, Y_i, Z_i \) are the coordinates of the world coordinate system.

Eliminating the parameter \( s_i \) of (1), simplified into

\[
\begin{align*}
  u_i &= \frac{m_4 X_i + m_5 Y_i + m_6 Z_i + m_7}{m_1 X_i + m_2 Y_i + m_3 Z_i + 1} \\
  v_i &= \frac{m_8 X_i + m_9 Y_i + m_{10} Z_i + m_{11}}{m_1 X_i + m_2 Y_i + m_3 Z_i + 1}
\end{align*}
\]

(2)

(3)

It is needed as least 6 feature points to construct an over determined equations. 11 parameters are solved using the least squares fitting method.

The visual sensor’s optical axis is perpendicular to the surface of the rotor, and intersects with the centre of the sphere. Firstly, accurately positioning the location of the rotor and stator, using the known coordinates of the spherical feature points in the world coordinate system to calibrate the sensor. Ideally rotor only rotates around the centre of the sphere and the sensor location is fixed after calibration, so the sensor calibration needs only once.

Let the coordinates of feature points in the world coordinate system and the rotation coordinate system can be expressed as \((X_i, Y_i, Z_i), (x_{i}, y_{i}, z_{i})\) \((i = 1, 2, \cdots n)\) respectively. The corresponding vectors are set as \( \hat{r}_{r,i} , \hat{r}_{i} \). The feature points distribute on the rotor surface and the constraint equation \( X_i^2 + Y_i^2 + Z_i^2 = r^2 \) (\( r \) denotes the spherical radius), combined with (2) and (3), the feature points on the image can be converted to coordinates in the world coordinates.

In photogrammetry, using the known coordinates of the points in two different coordinate systems, calculate the positional relationship between the coordinate system is a classic problem.
Reference [16] presented the closed solution to achieve absolute orientation quaternion. Set the coordinate system rotation matrix \( R \), the coordinate inter-relationship can be expressed as:

\[
\hat{r}_r = sR(\hat{r}_l) + \hat{r}_0
\]

where \( s \) is the proportional factor, \( r_0 \) is the origin point offset scale.

The principle of calculation matrix is to make \( \sum_{i=1}^{n} \left\| \hat{r}_{r,i} - sR(\hat{r}_{l,i}) - \hat{r}_0 \right\|^2 \) reach its minimal value, i.e. to make \( \sum_{i=1}^{n} \hat{r}_{r,i}' \cdot R(\hat{r}_{l,i}) \) to its maximal value.

Let \( \hat{r}_l = \frac{1}{n} \sum_{i=1}^{n} \hat{r}_{l,i} \), \( \hat{r}_r = \frac{1}{n} \sum_{i=1}^{n} \hat{r}_{r,i} \), \( \hat{r}_{r,i}' = \hat{r}_{r,i} - \hat{r}_l \), \( \hat{r}_{r,i}' = \hat{r}_{r,i} - \hat{r}_r \). Using the quaternion to represent the rotation \( \hat{r}' = \hat{q} \hat{q}' \), \( \hat{q} \) is the quaternion.

\[
\sum_{i=1}^{n} \hat{r}_{r,i}' \cdot R(\hat{r}_{l,i}) = \hat{q}^T N \hat{q}
\]

\[
N = \begin{bmatrix}
S_{xx} + S_{yy} + S_{zz} & S_{yz} - S_{zy} & S_{xz} - S_{zx} & S_{yx} - S_{xy} \\
S_{yx} - S_{xy} & S_{xx} - S_{yy} - S_{zz} & S_{xy} + S_{yx} & S_{zx} + S_{xz} \\
S_{xz} - S_{zx} & S_{xy} + S_{yx} & S_{zz} - S_{xx} - S_{yy} & S_{yz} + S_{zy} \\
S_{zy} - S_{yz} & S_{yx} + S_{xy} & S_{yy} - S_{xx} - S_{zz} & S_{xx} + S_{yy}
\end{bmatrix}
\]

where \( S_{xy} = \sum_{i=1}^{n} \hat{x}_{i,j} \cdot \hat{x}_{r,i}' \), \( \hat{q} \) is the eigenvector of the largest positive eigenvalue, \( \hat{q}' \) is the conjugate of \( \hat{q} \).

By solving above equation, it can be derived as

\[
s = \frac{\sum_{i=1}^{n} (\hat{q} \hat{r}_{l,i}' \hat{q}') \cdot \hat{r}_{r,i}'}{\sum_{i=1}^{n} \hat{r}_{r,i}' \cdot \hat{r}_{r,i}'}
\]

Substituting \( s \) into (4), based on the known feature points, the two sets of coordinate values under two coordinate systems, the over determined equations can be set to solve the rotation matrix \( R \) and the offset value \( \hat{r}_0 \).

The transformation matrix between the world coordinate and the rotation coordinate system can be written as
\[
C = \begin{pmatrix}
c \beta c \gamma & -c \beta s \gamma & s \beta \\
c \alpha s \gamma + s \alpha \beta c \gamma & c \alpha c \gamma - s \alpha s \beta s \gamma & -s \alpha c \beta \\
s \alpha s \gamma - c \alpha \beta c \gamma & c \alpha s \beta \gamma + s \alpha c \gamma & c \alpha c \beta
\end{pmatrix}
\]

(7)

where \(a, \beta, \gamma\) are the Euler angles, \(\cos\) and \(\sin\) denote the cosine and sine functions.

So \(\dot{r}_i = C \cdot \dot{r}_i\). When the centre of sphere is fixed, \(C = s R\). Three general Euler angles \(a, \beta, \gamma\) can be solved. The space positions of permanent magnets after rotor rotation are determined from the rotation matrix, so as to provide position feedback control signal.

b. Magnetic orientation sensor measurement of M-DOF motions

Magnetic sensors are widely available at low cost due to the rapid advancement of semiconductor fabrication technology. The Hall-effect sensors which can measure either a constant or a varying flux possess both small size and electronics elements compatibility. The distributed multi-pole (DMP) method can be effectively used for the multi-DOF actuator or motor torque modeling and other related magnetic field problems [17-20]. Unlike other field calculation methods (such as FEM), the DMP models can be computed in real-time for motion estimation. Once the magnetic fields are obtained, the forces and torques on the PM assembly can also be computed from the surface integration in terms of a Maxwell stress tensor. The Schematic illustration of the magnetic field sensors for orientation detection can be shown in figure 4.
Using the DMP modeling method, in the x-y-z coordinate system the closed-form solution characterizing the magnetic flux density $B$ of the single PM on the shaft is given by [18]

$$B = \frac{\mu_0}{4\pi} \sum_{j=0}^{k} m_j \sum_{i=1}^{n} \left( \frac{a_{Rji+}}{R_{ji+}^2} - \frac{a_{Rji-}}{R_{ji-}^2} \right)$$

where \( a_{Rji\pm} = \frac{P - P_{ji\pm}}{|P - P_{ji\pm}|^2} \), \( \mu_0 \) is the permeability of free space, \( m_j \) is the dipole moment in the \( j \)th loop; \( P - P_{ji\pm} \) is the position vector from the \( i \)th dipole in the \( j \)th loop to arbitrary position \( P(x, y, z) \) of interest; and \( R_{ji\pm} = |P - P_{ji\pm}| \).

The general flux density $B$ can be approximated as an $n$th order polynomial with the form:

$$\hat{B}_k = (\alpha, \beta) = \sum_{i=0}^{n} \sum_{j=0}^{n} c_{ij} C^i_k S^j_k$$

where the subscript \( k \) denotes the \( k \)th sensor; and \( n \) is the orders of the approximation.

After deriving $\hat{B}$, the orientation angle $(\alpha, \beta)$ can be solved as the inverse problem, which is related with the order of approximation. The following two equations can be used for each pair of sensors opposite on the axis:
\[
\begin{align*}
\bar{B}_A &= \frac{\hat{B}_{A+} + \hat{B}_{A-}}{2} = \sum_{i=0}^{n} \left( \sum_{j=0}^{n} c_{ij} C_{\alpha} \right) S_{\beta}^i, \text{ when } i \text{ is even} \\
\bar{B}_A &= \frac{\hat{B}_{A+} - \hat{B}_{A-}}{2} = \sum_{i=1}^{n} \left( \sum_{j=0}^{n} c_{ij} C_{\alpha} \right) S_{\beta}^i, \text{ when } i \text{ is odd}
\end{align*}
\]

where \( C \) and \( S \) denote the cosine and sine function of subscript angles respectively.

Then two sets of \((\alpha, \beta)\) estimating values can be solved from the pair of equations, one from each sensor pair.

c. Neural network based multisensor integration and fusion

The multisensory information fusion technology mainly aims to solve the information processing problems, which focuses on the using the information derived by multiple sensors to construct complete and proper description on the special object or environment features [21, 22]. Multisensor fusion and integration is a rapidly evolving research area and requires interdisciplinary knowledge in control theory, signal processing, artificial intelligence, probability and statistics, etc. There has been much research on the subject of multisensor and fusion in recent years. A number of researchers have reviewed the multisensor fusion algorithms, architectures, and applications.

During current neural networks’ development, the back propagation theory based feed-forward neural network (BP) is more mature and widely used as shown in figure 5 and parameters in table 1. Neural network is with the essence of the nonlinear system that can approximate any nonlinear relationship, which is very appropriate for the data fusion applications as in this work. From the above, it can be known that the orientation detection data by machine vision and magnetic field sensors methods can be integrated and fused. The work can be divided into two parts: firstly, the data derived from the machine vision measurement can be taken as training samples for the magnetic field sensors mapping relations constructing; secondly, in the case of single measurement method in effect. It is known that the PM actuator has \( x, y, z \) rotation angles in three different degrees-of-freedom, so it needs 4 inputs and 3 outputs, this paper uses a neural network to realize the presented fusion and calibration method.
Table 1: Parameters of the neural network for applications

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Parameter Value</th>
<th>Parameter</th>
<th>Parameter Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of Input Neural Nodes</td>
<td>4</td>
<td>Minimum Mean Squared Error</td>
<td>1e-1</td>
</tr>
<tr>
<td>Number of output Neural Nodes</td>
<td>3</td>
<td>Minimum Gradient</td>
<td>1e-20</td>
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<tr>
<td>Hidden Layer Neural Nodes' Transfer Function</td>
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<td>Maximum Training Times</td>
<td>500</td>
</tr>
<tr>
<td>Output Neural Nodes’ Transfer Function</td>
<td>purelin</td>
<td></td>
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</tr>
</tbody>
</table>

Figure 5. The structure of BP neural network

After calibration, certain sets of motion measurement data can be collected and used as the input of neural network with 120 sets as the training samples and 60 sets as the test samples. The number of hidden nodes is determined by the empirical formula $m = \sqrt{n + l + \alpha}$, where $m$ denotes the number of hidden nodes, $n$ denotes the number of input layer nodes, $l$ denotes the number of output layer nodes, and $\alpha$ is constant number of 1~10. According to the input and output nodes, the nodes’ numbers of hidden layer are set with 6, 7, 8, 9, 10, 11, and 12 to train and test. The training process can be shown in figure 6.
IV. RESULTS AND DISCUSSION

In order to verify the reliability of the presented orientation detection methodology of combined two modes, an experiment platform was built and applied for this work. The spherical rotor and grid painting parameters: the width of the grid lines along the spherical angle is 0.4 degree. The sensor optical axis is directed through the center of the sphere. In the initial rotor position, the world coordinate system is fixed and coincided with the rotor moving coordinate system. Due to the limits of experimental conditions, in the experiment, only the upper hemisphere of rotor is used and the feature points are directly marked on the rotor sphere surface. The experiment device can be shown in figure 7. The machine vision system contains the NI CVS vision disposal system and the main vision sensor uses the SONY XCD-X710 CCD camera. The manufactured experiment actuator prototype can be shown in figure 8. Four Hall Effect sensors are taken as the magnetic field detection element. Detailed Hall sensor is shown in figure 9.

Calibrate the vision sensor at the initial position and the vision sensor matrix can be derived as

\[
M = \begin{bmatrix}
-536.47 & -2763.7 & 3090.1 & 351.43 \\
-3806.6 & -1131 & -1104.3 & 396.11 \\
-0.036092 & -3.2893 & -1.3373 & 1
\end{bmatrix}
\]
Figure 7. The photograph of experimental platform

Figure 8. Experiment prototype of multi-DOF actuator

Figure 9. Experiment built Hall Effect sensor

The experiment motion is set with three tilt rotations in sequence, with single rotation with respect to X axis, Y axis and both combined rotation. The measurement data by the two modes are collected and compared. The real standard motion data was firstly implemented by high precision optical encoder for reference with repeated processes. The measured trajectory data can be shown in figure 10. Figure 11 gives the errors of different measurement modes for comparison. To test the reliability and practicability, the fusion effects in the case of magnetic sensor fault are evaluated. The magnetic sensor is exerted with an outer disturbance at the interface of the first and second segment as shown in figure 10 and 11 of sudden change on the trajectory and error data. It can be seen from the results, compared with single machine vision and magnetic sensor method, the neural network based data fusion application can improve the tracking accuracy with lower errors and anti-interference ability.
V. CONCLUSIONS

This paper presents the multisensor fusion based orientation detection methodology for multi-DOF actuator or motor applications. The two modes as machine vision and magnetic field sensor are all applied in related industrial areas with their special features of advantages. In this work,
the calibrated machine vision mode is used for neural network training and magnetic field sensor modeling. Experiment is carried out to validate the effectiveness of presented methodology by setting three tilt motion segments in sequence. The results prove the methodology is available and effective. It is expected that the presented scheme is able to provide a reference for further research and investigation for similar measurement problems of multi-DOF actuators or motors.

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