

## Study on the Structural Parameter Coupling of Articulated Arm Coordinate Measuring Machines

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### Abstract

The articulated arm coordinate measuring machine (AACMM) is a kind of new coordinate measuring device based on non-Cartesian system. The measuring accuracy of the AACMM can be effectively improved by parameter identification. However, some of the structural parameters are coupling (linearly related), so the structural parameters usually cannot be identified correctly. The Jacobian matrix was obtained through differential transformation of the kinematic model of a 6-DOF AACMM. The rank calculation results show that the Jacobian matrix is not full rank, which means some structural parameters are linearly related. Singular value decomposition and the elementary row transform of the Jacobian matrix were carried out to determine the linearly related structural parameters. And two pairs of structural parameters were found to be linearly related, which can't be identified at the same time. Finally, the linear correlation was applied in the structural parameters identification, and the results show that the linear correlation can make the identification obtain correct structural parameters.

**Keywords:** articulated arm coordinate measuring machine, structural parameter coupling, kinematic model, parameter identification, Jacobian matrix

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### 1. Introduction

The AACMM is a multi-DOF (typically 6-DOF) and non-Cartesian coordinate measuring machine (CMM), which is modeled according to the structure of human joints: a series of linkages connected by rotating joints. Comparing with traditional CMMs the AACMM has the features of small size, light weight, large measurement range, flexible and can be applied in industrial site [1, 2]. With these unique advantages the AACMM has been applied in the field of mold design, product quality online testing, equipment maintenance and assembly [3].

The accuracy of structural parameters is the main influencing factor to the measurement accuracy of AACMMs [4], the structural parameter identification is one of the main measures to improve the accuracy of AACMMs. Calibration is an integrated process of four steps including modeling, measurement, parameter identification and error compensation [5, 6]. Selection of the appropriate kinematics model and the calibration model is the premise of calibration of AACMM, and processing the data of calibration to realize structural parameter identification so as to achieve the calibration of AACMMs. In parameter identification linearly related structural parameters can resulting Jacobian matrix to be singular, and the solution is not the required structural parameters we need [7-9]. So the linearly related structural parameters must be determined before identification.

### 2. The Kinematic Model

To study the relationship between the coordinate of probe and rotary angle of every joint Denavit-Hartenberg (D-H) method was used to model the measuring machine coordinate system, as shown in Figure 1 and Figure 2. According to D-H method the coordinate transformation matrix between the adjacent coordinate systems is shown in Eq. 1 and the nominal structural parameters of the AACMM is shown in Table 1.

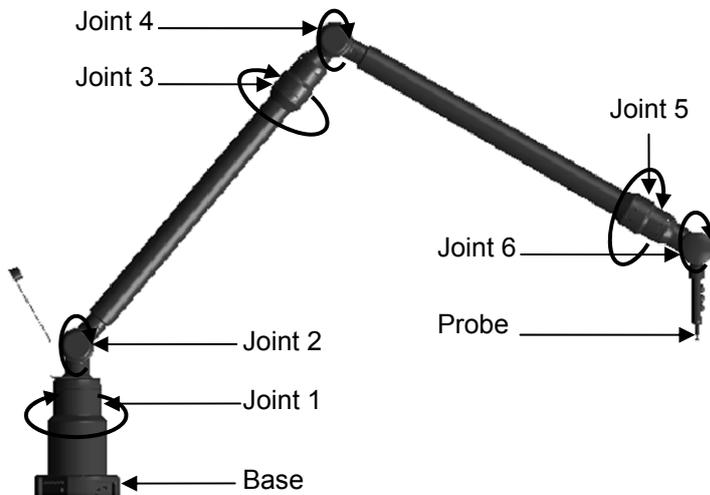


Figure. 1 The structure of the AACMM

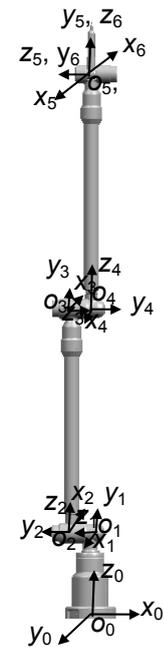


Figure. 2 The coordinate systems of the AACMM

$$T_{i-1,i} = \begin{bmatrix} \cos \theta_i & -\sin \theta_i \cos \alpha_i & \sin \theta_i \sin \alpha_i & a_i \cos \theta_i \\ \sin \theta_i & \cos \theta_i \cos \alpha_i & -\cos \theta_i \sin \alpha_i & a_i \sin \theta_i \\ 0 & \sin \alpha_i & \cos \alpha_i & d_i \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (1)$$

Where  $d_i$  is joint offset,  $a_i$  is link length,  $\alpha_i$  is twist angle,  $\theta_i$  is joint angle

Table 1. The nominal value of the structural parameters of the AACMM

Linkage No.	$a_i$ [mm]	$d_i$ [mm]	$\theta_i$ [°]	$\alpha_i$ [°]	$l$ [mm]
1	0	300	$\theta_1$	90	183.5
2	0	65.5	$\theta_2$	90	
3	0	792.7	$\theta_3$	90	
4	0	65.5	$\theta_4$	90	
5	0	883.2	$\theta_5$	90	
6	0	0	$\theta_6$	90	

The coordinates of the probe in the coordinate of  $\{x_6y_6z_6\}$  are  $(l_x, l_y, l)$ . Because the probe is equivalent to the axis of  $z_6$  of translational transformation in  $\{x_6y_6z_6\}$ , its coordinates in  $\{x_6y_6z_6\}$  should be  $(0,0,l)$  and its coordinates in  $\{x_0y_0z_0\}$  can be calculated through Eq. 2.

$$P = T_{0,1} \cdot T_{1,2} \cdot T_{2,3} \cdot T_{3,4} \cdot T_{4,5} \cdot T_{5,6} \cdot [0, 0, l, 1]^T \quad (2)$$

### 3. Analysis of Structural Parameter Coupling

#### 3.1. Error Model of the AACMM

In practical applications, due to the presence of manufacturing and assembling error, the change of environmental temperature and force-deformation the structural parameters will generate errors.  $\Delta a_i$  and  $\Delta d_i$  are the linkage length errors generated by manufacturing and assembly.  $\Delta \alpha_i$  is the angle error produced by adjacent axis parallelism and perpendicularity.  $\Delta \theta_j$  is the zero offset error generated by the angle of encoder zero and nominal model joint rotation

zero misalignment in the assembly process. Considering each coordinate system has four structural parameters and the probe length there are totally 25 structural parameter errors should be calibrated.

When we specify the probe of AACMM moves to the position  $P$ , in fact because of these structural parameter errors the probe will move to the actual position  $R$ , the position errors will be:

$$\Delta P = R - P \quad (3)$$

If these structural parameters are small enough we can use differential kinematics model approximation instead of error equation [9, 10]. The probe position error of the AACMM by total differential of the kinematics can be written as Eq. 4.

$$\Delta P = \sum_{i=1}^6 \frac{\partial P}{\partial \theta_i} \Delta \theta_i + \sum_{i=0}^5 \frac{\partial P}{\partial \alpha_i} \Delta \alpha_i + \sum_{i=0}^5 \frac{\partial P}{\partial a_i} \Delta a_i + \sum_{i=1}^6 \frac{\partial P}{\partial d_i} \Delta d_i + \frac{\partial P}{\partial l} \Delta l \quad (4)$$

Eq. 4 is the error model of the AACMM.

### 3.2. Analysis of Structural Parameter Coupling

Eq. 4 is written in matrix form as shown in Eq. 5.

$$\Delta \mathbf{P} = \mathbf{J} \Delta \mathbf{Q} \quad (5)$$

Where  $\Delta \mathbf{P}$  is a matrix of  $3 \times 1$ ,  $\Delta \mathbf{Q}$  is a matrix of  $25 \times 1$ , the Jacobin matrix  $\mathbf{J}$  is a matrix of  $3 \times 25$ . To solve the value of the 25 structural parameters' error in  $\Delta \mathbf{Q}$ ,  $N$  ( $\geq 9$ ) measuring points must be selected to constitute over-determined equations. However, in practical operation there are linearly related structural parameters which can make Jacobian matrix  $\mathbf{J}$  to be singular. And the solution obtained from the equations is not the required value we need. Therefore the linearly related structural parameters must be found out and eliminated from the identification list. In matrix theory linearly related row can be found by singular value decomposition elementary row transformation of the orthogonal matrix decomposition [11, 12]. Premultiplying  $\mathbf{J}^T$  for both sides of Eq.5 can obtain Eq. 6.

$$[\mathbf{J}^T \mathbf{J}] \Delta \mathbf{Q} = \mathbf{J}^T \Delta \mathbf{P} \quad (6)$$

Let  $\mathbf{H} = [\mathbf{J}^T \cdot \mathbf{J}]$ , then:

$$\mathbf{H} = \mathbf{U} \begin{bmatrix} \mathbf{S} & 0 \\ 0 & 0 \end{bmatrix} \mathbf{V}^T \quad (7)$$

Where  $\mathbf{U}$  and  $\mathbf{V}$  are orthogonal matrix of  $25 \times 25$ ,  $\mathbf{S} = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_r)$  ( $r \leq 25$ ),  $r$  is the rank of the matrix  $\mathbf{H}$  and the Jacobian matrix  $\mathbf{J}$ . So the number of linearly related parameters is  $25-r$ . From Eq. 7 and Eq. 6 we can obtain Eq. 8.

$$\mathbf{V}^T \Delta \mathbf{Q} = \begin{bmatrix} \mathbf{S}^{-1} & 0 \\ 0 & 0 \end{bmatrix} \mathbf{U}^{-1} \mathbf{J}^T \Delta \mathbf{P} \quad (8)$$

$\mathbf{H}$  is a symmetric matrix,  $\mathbf{V}^T = \mathbf{U}^{-1}$  and  $\mathbf{V}$  is rotation matrixes, so  $\mathbf{V}^T \cdot \Delta \mathbf{Q}$  is equivalent to  $\Delta \mathbf{Q}$ . The linearly related structural parameters in  $\Delta \mathbf{Q}$  can be found out by elementary row transformation of the last 4 lines, as shown in Eq. 9.

$$\Delta \mathbf{P} = \mathbf{J1} \Delta \mathbf{Q1} \quad (9)$$

Where  $\mathbf{J1}$  is a matrix of  $(3 \times N) \times r$ ,  $\Delta \mathbf{Q1}$  is a matrix of  $r \times 1$ ,  $\Delta \mathbf{P}$  is the matrix of  $(3 \times N) \times 1$ .

The rank of  $\mathbf{J}$  is 23, thus there are two linearly related structural parameters. We can obtain linearly related parameters by carrying out singular value decomposition and the elementary row transform.

$$\Delta a_6 = l\Delta\theta_6 \quad (10)$$

$$\Delta d_6 = l\Delta\alpha_6 \quad (11)$$

As shown in Eq. 10 and 11  $a_6$  is linearly related with  $\theta_6$  and  $d_6$  is linearly related with  $\alpha_6$ .

#### 4. Test of Structural Parameter Coupling in Identification

We assume there are some errors in the structural parameters, as shown in Table 2. These errors can result in measuring error of AACMMs.

Table 2. The set errors of the structural parameters

Linkage No.	$\Delta a_i$ [mm]	$\Delta d_i$ [mm]	$\Delta\theta_i$ [°]	$\Delta\alpha_i$ [°]	$\Delta l$ [mm]
1	0.10	-0.05	0.0014	0.0010	-0.03mm
2	-0.08	0.11	-0.0023	0.0014	
3	-0.12	0.06	0.0017	-0.0012	
4	0.03	0.13	0.0010	0.0017	
5	0.07	-0.04	-0.0014	0.0010	
6	0	0	0.0010	-0.0010	

To identify the errors of the structural parameters by the method proposed in this paper, we need ten groups of joint angles which were given in Table 3.

Table 3. The joint angles used in identification

No.	1	2	3	4	5	6	7	8	9	10
$\theta_1$ [°]	0	180	180	90	90	90	0	90	120	0
$\theta_2$ [°]	180	120	90	180	180	180	150	-90	180	-90
$\theta_3$ [°]	0	0	90	90	90	-90	-90	0	-90	-180
$\theta_4$ [°]	180	180	90	90	180	90	90	-90	120	-90
$\theta_5$ [°]	0	0	0	90	90	-90	90	0	90	0
$\theta_6$ [°]	180	180	180	0	90	90	-90	-90	60	-90

As  $a_6$  is linearly related with  $\theta_6$  and  $d_6$  is linearly related with  $\alpha_6$ , the errors of  $a_6$  and  $d_6$  can be compensated by  $\theta_6$  and  $\alpha_6$ . In the parameter identification the errors of  $a_6$  and  $d_6$  are set to be 0. With the data in Table 3 the least squares solution of Eq. 9 was reached, as shown in Table 4.

Table 4. The identified errors of the structural parameters

Linkage No.	$\Delta a_i$ [mm]	$\Delta d_i$ [mm]	$\Delta\theta_i$ [°]	$\Delta\alpha_i$ [°]	$\Delta l$ [mm]
1	0.100	-0.052	0.0014	0.0010	-0.035mm
2	-0.078	0.098	-0.0023	0.0014	
3	-0.126	0.060	0.0017	-0.0012	
4	0.045	0.131	0.0010	0.0017	
5	0.068	-0.046	-0.0014	0.0010	
6	0	0	0.0010	-0.0010	

Table 4 and Table 2 shows that the identified errors are all nearly equal to the set errors, the mean and the maximum difference of  $\Delta a_i$ ,  $\Delta d_i$  and  $\Delta l$  is 0.005mm and 0.015mm, the difference of  $\Delta\theta_i$  and  $\Delta\alpha_i$  is 0°.

Structural parameter identification can improve the measurement accuracy of AACMMs. Table 5 shows the measurement errors before and after identification of the AACMM.

Table 5. Measurement errors before and after identification of the AACMM

No.	Error Before Identification[mm]	Error After Identification[mm]
1	(5.457, -2.687, -0.086)	(-0.006, -0.011, 0.014)
2	(-2.751, -0.498, -4.704)	(-0.007, 0.010, 0.013)
3	(-4.434, -2.144, -6.955)	(-0.016, 0.022, 0.005)
4	(-0.144, 5.075, -5.824)	(0.013, -0.007, -0.03)
5	(-1.276, 6.517, 1.256)	(0.012, 0.002, 0.011)
6	(0.455, -2.055, 0.196)	(-0.007, -0.001, -0.002)
7	(-1.900, -0.198, 1.025)	(-0.002, 0.006, -0.001)
8	(1.008, 3.103, 1.239)	(0.002, -0.008, -0.006)
9	(-0.284, -0.946, 0.610)	(-0.007, -0.002, -0.001)
10	(-0.962, 0.783, 2.178)	(-0.003, 0.007, -0.002)

Table 5 shows that before structural parameter identification the maximum error is 6.955mm while after identification the maximum error is 0.022mm. So the measurement accuracy of the AACMM was improved greatly.

#### 4. Conclusion

The kinematic model and coordinate systems based on D-H model were established. The rank calculation results show that the Jacobian matrix is not full rank, which means some structural parameters are linearly related. Singular value decomposition and the elementary row transform of the Jacobian matrix were carried out to determine the linearly related structural parameters. And that  $a_6$  is linearly related with  $\theta_6$  and  $d_6$  is linearly related with  $\alpha_6$  were found to be linearly related, which can't be identified at the same time. Finally, the linear correlation was applied in the structural parameters identification, and the results show that after identification the position error of the AACMM was improved greatly.

#### Acknowledgements

This work was financially supported by the Yunnan Provincial Natural Science Foundation of China (Grant No.2011FB028).

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