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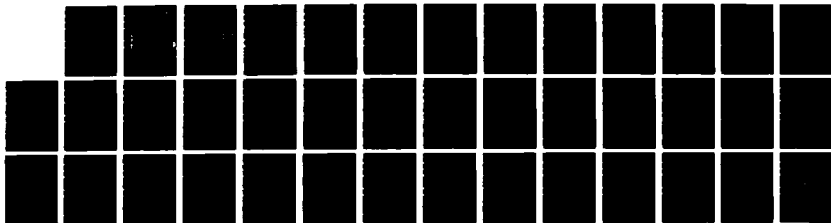
CAMERA CALIBRATION(U) ROCHESTER UNIV NY DEPT OF  
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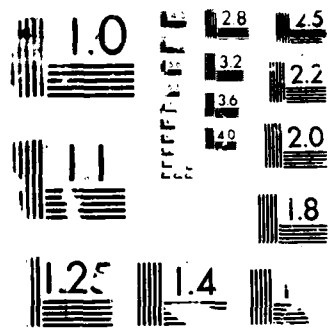
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Camera Calibration

Isidore Rigoutsos and Christopher M. Brown  
Department of Computer Science  
The University of Rochester  
Rochester, NY 14627

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March 1986

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## Camera Calibration

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### Abstract

Given an image that has been scaled both horizontally and vertically (possibly with different scale factors in the two directions), we determine the camera position and orientation, as well as the scale factors for sampling in the  $u$  and  $v$  axes of the image plane, from the knowledge of the correspondence between a few known points in space and their location on the image (image plane).

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### **Abstract**

Given an image that has been scaled both horizontally and vertically ( possibly with different scale factors in the two directions ), determine the camera position and orientation, as well as the scale factors for sampling in the  $u$  and  $v$  axes of the image plane, from the knowledge of the correspondences between a few known points in space and their locations on the image ( image plane ).

### **Introduction**

For many practical purposes, it is sufficient to model the imaging process of a camera by a pure point projection process; then the two-dimensional image plane coordinates of a 3-D point, as seen by a camera, can be easily expressed in terms of a  $3 \times 4$  matrix using the notion of the homogeneous coordinate system[2,5]. Because of its elegance and its usual accuracy, this matrix has considerable applications in fields like computer graphics, computer vision, motion tracking, etc. The problem of camera calibration and determination of the above mentioned matrix has received considerable attention in the literature[3,4,5,6]. Camera calibration is simply determining the elements of the matrix that describes the imaging process. Equivalently, it can be thought of as determining certain geometrical and optical constants, such as the direction of the optical axis, the spatial location of the camera's nodal point and the focal length of the lens. The solution that is described in what follows, is the one suggested by Ganapathy ( [5] ).

## 1. Formulation of the problem.

Consider a right handed three-dimensional orthonormal coordinate system OXYZ ( see Fig. 1 ). Using the homogeneous coordinate system ( [ 2 ] ) the coordinates of a point A( x, y, z ) are represented by a quadruple of the form ( ax, ay, az, a ) where we introduced the scalar a . In order to obtain the three-dimensional coordinates of a given point that is represented in the homogeneous coordinate system, one merely divides by the last component of the quadruple. In a similar way, a ( two-dimensional ) point I ( u, v ) of the image plane ( U-V ) is represented by the triple (iu, iv, i ), for some arbitrary scalar quantity i, and we divide by the third component i to obtain u and v.

Consider a camera having its center at C that is located at ( x<sub>0</sub>, y<sub>0</sub>, z<sub>0</sub> ) ( see Fig.1); the camera looks along a line of sight ( optical axis of the camera ) CO'P where P is the point at which the optical axis of the camera intersects the X-Y plane and O' the point at which the optical axis pierces the image plane. Taking a right-handed orthonormal coordinate system O''X'Y'Z' centered at C, we can write for any point of the 3-D space the following equation:

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} + \begin{bmatrix} p \\ q \\ r \end{bmatrix} \quad (1)$$

where

$$R = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$

is the rotation matrix, and

$$\begin{bmatrix} p \\ q \\ r \end{bmatrix}$$

is the translation matrix.

In the above,  $(x, y, z)$  and  $(x', y', z')$  are the coordinates with respect to the OXYZ and O'X'Y'Z' coordinate systems respectively.

Using homogeneous coordinates, the above equation (1) can be written as follows:

$$\begin{bmatrix} jx' \\ jy' \\ jz' \\ j \end{bmatrix} = \begin{bmatrix} a & b & c & p \\ d & e & f & q \\ g & h & i & r \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} lx \\ ly \\ lz \\ 1 \end{bmatrix} \quad (2)$$

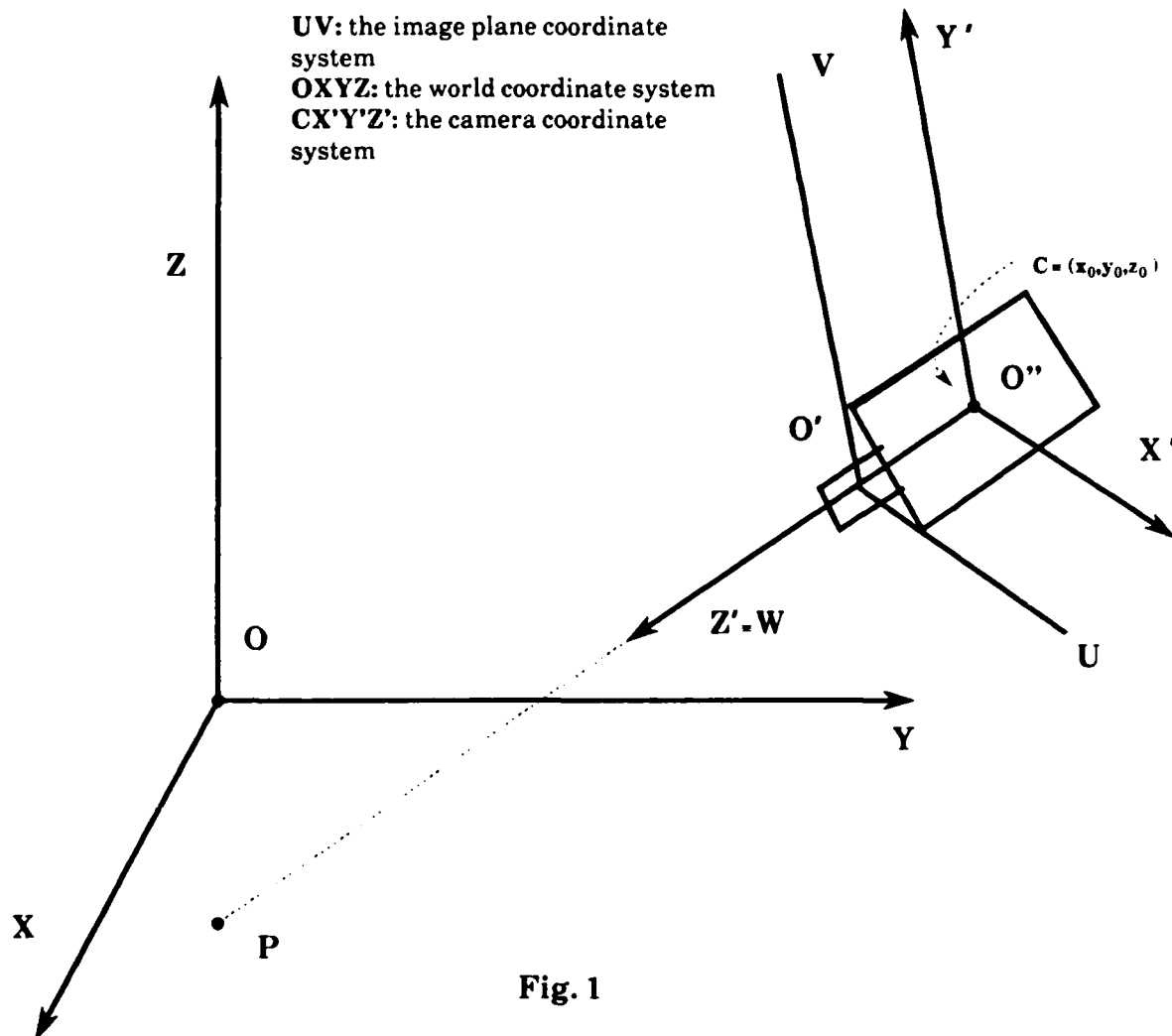
where we combined both the translation and rotation matrices in one.

Since the  $(x_0, y_0, z_0)$  point of the OXYZ orthonormal system is mapped to the  $(0, 0, 0)$  point of the O'UVW system, then using equation (1) it follows that:

$$p = -a * x_0 - b * y_0 - c * z_0 \quad (3)$$

$$q = -d * x_0 - e * y_0 - f * z_0 \quad (4)$$

$$r = -g * x_0 - h * y_0 - i * z_0 \quad (5)$$



Substituting into equation (1)  $(x, y, z)$  by  $(x_0, y_0, z_0)$  and  $(x', y', z')$  by  $(0, 0, 0)$  and solving for the translation matrix, we have the following equation:

$$\begin{bmatrix} -p \\ -q \\ -r \end{bmatrix} = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix} \quad (6)$$

But R, the rotation matrix, is assumed to be an orthonormal rotation matrix and therefore it follows that  $R^{-1} = R^t$ . Using the latter equation we can write (6) as:

$$\begin{bmatrix} a & d & g \\ b & e & h \\ c & f & i \end{bmatrix} \begin{bmatrix} -p \\ -q \\ -r \end{bmatrix} = \begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix} \quad (7)$$

This equation can be equivalently written as :

$$x_0 = -p * a - q * d - r * g \quad (8)$$

$$y_0 = -p * b - q * e - r * h \quad (9)$$

$$z_0 = -p * c - q * f - r * i \quad (10)$$

Consider now a point  $(x', y', z')$  where the coordinates are taken with respect to the  $O'X'Y'Z'$  system. The image plane coordinates of the considered point will therefore be given by the equations:

$$u = \frac{fl * x'}{z'} \quad \text{and} \quad v = \frac{fl * y'}{z'}$$

Here we assumed that the  $Z'$  axis is parallel to the optical axis and that the image plane is in front of the camera;  $fl$  is the focal length of the camera. Furthermore,  $X', Y'$ , and  $Z'$  axes can be aligned with the  $U, V$ , and  $W$  axes such that  $U$ -axis is parallel to the  $X'$ -axis,  $V$ -axis is parallel to the  $Y'$ -axis, and  $W$ -axis is parallel to the  $Z'$ -axis. In practice though, the measurements in both the  $X'$  and  $Y'$  axes are subject to finite resolution; this consequently induces a discrete spectrum of values for the measured quantities. To be general we will assume two scale factors (possibly different)  $k_u$  and  $k_v$  for sampling in the  $X'$  and  $Y'$  axes respectively. Moreover, if  $(u_0,$

$v_0, 0$ ) are the coordinates, with respect to the UVW coordinate system, of the point that we consider to be the one through which the optical axis passes, then the above equations can be written as:

$$u = u_0 + \frac{k_u * fl * x'}{z'} \quad (11) \quad \text{and} \quad v = v_0 + \frac{k_v * fl * y'}{z'} \quad (12)$$

Representing now the products  $k_u * fl$  and  $k_v * fl$  by  $k_1$  and  $k_2$  respectively, we can clearly write the above equations, using homogeneous coordinates, as follows:

$$\begin{bmatrix} wu \\ wv \\ w \end{bmatrix} = \begin{bmatrix} k_1 & 0 & u_0 & 0 \\ 0 & k_2 & v_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} sx' \\ sy' \\ sz' \\ s \end{bmatrix} \quad (13)$$

or

$$\begin{bmatrix} wu \\ wv \\ w \end{bmatrix} = M_1 \begin{bmatrix} sx' \\ sy' \\ sz' \\ s \end{bmatrix} \quad (14)$$

Note that  $M_1$  is a 3x3 matrix whose determinant is equal to  $k_1 * k_2 \neq 0$ . Therefore it is invertible.

Similarly, equation (2) can be written as:

$$\begin{bmatrix} sx' \\ sy' \\ sz' \\ s \end{bmatrix} = M_2 \begin{bmatrix} lx \\ ly \\ lz \\ 1 \end{bmatrix} \quad (15)$$

Observe that  $M_2$  is completely determined by the camera location and orientation. Equations ( 14 ) and ( 15 ) can be written as :

$$\begin{bmatrix} wu \\ wv \\ w \end{bmatrix} = M_3 \begin{bmatrix} sx \\ sy \\ sz \\ s \end{bmatrix} \quad (16)$$

where  $M_3 = M_1 \cdot M_2$  is a 3x4 matrix that is known as the calibration matrix. Let us denote its components by  $t_i / i = 1,2,3,4, \dots, 12$  , where:

$$t_1 = k_1 * a + u_0 * g$$

$$t_2 = k_1 * b + u_0 * h$$

$$t_3 = k_1 * c + u_0 * i$$

$$t_4 = k_1 * p + u_0 * r$$

$$t_5 = k_2 * d + v_0 * g \quad \text{equations ( 17 )}$$

$$t_6 = k_2 * e + v_0 * h$$

$$t_7 = k_2 * f + v_0 * i$$

$$t_8 = k_2 * q + v_0 * r$$

$$t_9 = g$$

$$t_{10} = h$$

$$t_{11} = i$$

$$t_{12} = r$$

Since we use the homogeneous coordinates formulation, no change will occur if we divide all the  $t_i$ 's by  $t_{12}$ . Using the notation  $T_i \equiv t_i / t_{12}$ , we have:

$$T_1 = (k_1 * a + u_0 * g) / r$$

$$T_2 = (k_1 * b + u_0 * h) / r$$

$$T_3 = (k_1 * c + u_0 * i) / r$$

$$T_4 = (k_1 * p + u_0 * r) / r$$

$$T_5 = (k_2 * d + v_0 * g) / r \quad \text{equations ( 18 )}$$

$$T_6 = (k_2 * e + v_0 * h) / r$$

$$T_7 = (k_2 * f + v_0 * i) / r$$

$$T_8 = (k_2 * q + v_0 * r) / r$$

$$T_9 = g / r$$

$$T_{10} = h / r$$

$$T_{11} = i / r$$

$$T_{12} = 1$$

Therefore equation ( 16 ) can be expressed in terms of the eleven  $T_i$  's. Each observation ( u, v ) together with the corresponding ( x, y, z ) gives us two linear equations in the eleven  $T_i$  's. Hence a minimum of six observations is needed to determine the entries of  $M_3$ .

## 2. Some useful properties.

Consider again matrix R ( rotation matrix ). Since the latter is an orthonormal rotation matrix the following identities hold:

$$a^2 + b^2 + c^2 = 1 \quad (19)$$

$$d^2 + e^2 + f^2 = 1 \quad (20)$$

$$g^2 + h^2 + i^2 = 1 \quad (21)$$

$$d * g + e * h + f * i = 0 \quad (22)$$

$$a * d + b * e + c * f = 0 \quad (23)$$

$$a * g + b * h + c * i = 0 \quad (24)$$

Equations (19)-(21) represent the normality of the vectors whereas equations (22) - (24) represent the orthogonality of the vectors that correspond to the rows of the rotation matrix ([7]). Recalling that  $\text{Det}(R) = \text{Det}(R^{-1}) = 1$  and that  $R^{-1} = R^t$ , we have the following set of equations :

$$a = e * i - f * h \quad (25)$$

$$b = f * g - d * i \quad (26)$$

$$c = d * h - e * g \quad (27)$$

$$d = c * h - b * i \quad (28)$$

$$e = c * g - a * i \quad (29)$$

$$f = b * g - a * h \quad (30)$$

$$g = b * f - e * c \quad (31)$$

$$h = c * d - a * f \quad (32)$$

$$i = a * e - b * d \quad (33)$$

### 3. Solving the problem

The equations (19) through (33) we described above are not all independent. As a matter of fact the existing interdependence can be expressed by the following identity:

$$(a*d + b*e + c*f)^2 = (a^2 + b^2 + c^2)*(d^2 + e^2 + f^2) - (a*e - b*d)^2 - (c*d - a*f)^2 - (b*f - e*c)^2$$

Consider now the vectors **A**, **B**, and **C** where **A**, **B**, and **C** are defined as follows:

$$\mathbf{A} \equiv (T_1, T_2, T_3) \quad (34)$$

$$B \equiv (T_5, T_6, T_7) \quad (35)$$

$$C \equiv (T_9, T_{10}, T_{11}) \quad (36)$$

Using the definitions of the  $T_i$ 's, and after some easy computations we have that:

$$A \cdot A = \frac{k_1^2 + u_0^2}{r^2} \quad (37)$$

$$B \cdot B = \frac{k_2^2 + v_0^2}{r^2} \quad (38)$$

$$C \cdot C = \frac{1}{r^2} \quad (39)$$

$$A \cdot C = \frac{u_0}{r^2} \quad (40)$$

$$B \cdot C = \frac{v_0}{r^2} \quad (41)$$

The above equations clearly indicate the method that is to be used in order to solve the problem.

### 3 a) Computing the parameters of the problem

We first determine the entries  $T_i, i = 1, 2, 3, 4 \dots 11$  (recall that  $T_{12} = 1$ ). Then using the definitions ( 34 ), ( 35 ), ( 36 ) the equations ( 37 ) thru ( 41 ) and equations ( 18 ), we can clearly see that:

$$r = \frac{1}{\sqrt{(T_9^2 + T_{10}^2 + T_{11}^2)}} \quad (42)$$

$$u_o = (T_1 * T_9 + T_2 * T_{10} + T_3 * T_{11}) * r^2 \quad (43)$$

$$v_o = (T_5 * T_9 + T_6 * T_{10} + T_7 * T_{11}) * r^2 \quad (44)$$

$$k_1 = \sqrt{(r^2 * (T_1 * T_1 + T_2 * T_2 + T_3 * T_3) - u_o^2)} \quad (45)$$

$$k_2 = \sqrt{(r^2 * (T_5 * T_5 + T_6 * T_6 + T_7 * T_7) - v_o^2)} \quad (46)$$

$$g = T_9 * r \quad (47)$$

$$h = T_{10} * r \quad (48)$$

$$i = T_{11} * r \quad (49)$$

$$a = \frac{T_1 * r - u_o * g}{k_1} \quad (50)$$

$$b = \frac{T_2 * r - u_o * h}{k_1} \quad (51)$$

$$c = \frac{T_3 * r - u_o * i}{k_1} \quad (52)$$

$$d = \frac{T_5 * r - v_o * g}{k_2} \quad (53)$$

$$e = \frac{T_6 * r - v_o * h}{k_2} \quad (54)$$

$$f = \frac{T_7^* r - v_0^* i}{k_2} \quad (55)$$

$$p = \frac{T_4^* r - u_0^* r}{k_1} \quad (56)$$

$$q = \frac{T_8^* r - v_0^* r}{k_2} \quad (57)$$

In the above the sign of  $r$  is assumed to be positive. As a matter of fact the solutions with  $+r$  and  $-r$  correspond to the image in front of and behind the camera respectively ( see [5] ). Moreover,  $k_1$  can always be assumed positive. This is not true for  $k_2$  though. The sign of  $k_2$  depends on the polarity of axes  $U$  and  $V$  ( see [5] ). We can determine the sign of  $k_2$  using the set of equations  $\{(28), (29), (30)\}$  and  $\{(53), (54), (55)\}$ . Observe from the latter set of equations that the signs of  $d$ ,  $e$  and  $f$  depend on the sign of  $k_2$ ; therefore, the former set of equations will hold only for one of the signs of  $k_2$ , that is for the correct one.

### 3 b) How to determine the table entries

Observe that we deal with a total of 11 unknown table entries and therefore, in principle, a set of at least 5 1/2 points is needed. Since, in general, more than 6 points are used, one needs to apply the method of least-squares. Recalling the fact that we are using the homogeneous coordinate representation, we can clearly see that the  $u$  and  $v$  coordinates are given by the following equations:

$$u_i = \frac{T_1^* x_i + T_2^* y_i + T_3^* z_i + T_4}{T_9^* x_i + T_{10}^* y_i + T_{11}^* z_i + T_{12}} \quad / i = 1, 2, 3 \dots n \quad (58)$$

$$v_i = \frac{T_5^* x_i + T_6^* y_i + T_7^* z_i + T_8}{T_9^* x_i + T_{10}^* y_i + T_{11}^* z_i + T_{12}} \quad / i = 1, 2, 3 \dots n \quad (59)$$

where  $s, w$  were assumed to be equal to 1 ( see equation 16 ) and  $n$  is the number of the experimental points. Therefore, application of the least-squares method to this case will require that we minimize the quantity  $A$ , where:

$$A \equiv \sum_{i=1}^n [(T_9^* x_i + T_{10}^* y_i + T_{11}^* z_i + 1) * u_i - (T_1^* x_i + T_2^* y_i + T_3^* z_i + T_4)]^2 +$$

$$\sum_{i=1}^n [(T_9^* x_i + T_{10}^* y_i + T_{11}^* z_i + 1) * v_i - (T_5^* x_i + T_6^* y_i + T_7^* z_i + T_8)]^2$$

Taking the partial derivatives of  $A$  with respect to  $T_i$ , for  $i = 1, 2, 3, \dots, 11$ , and equating them to 0, we come up with the following set of linear equations :

$$\begin{aligned}
& T_1^* \left( \sum_{i=1}^n x_i^2 * u_i \right) + T_2^* \left( \sum_{i=1}^n x_i * y_i * u_i \right) + T_3^* \left( \sum_{i=1}^n x_i * z_i * u_i \right) + T_4^* \left( \sum_{i=1}^n x_i * u_i \right) + \\
& + T_5^* \left( \sum_{i=1}^n x_i^2 * v_i \right) + T_6^* \left( \sum_{i=1}^n x_i * y_i * v_i \right) + T_7^* \left( \sum_{i=1}^n x_i * z_i * v_i \right) + T_8^* \left( \sum_{i=1}^n x_i * v_i \right) - \\
& - T_9^* \left( \sum_{i=1}^n x_i^2 * (u_i^2 + v_i^2) \right) - T_{10}^* \left( \sum_{i=1}^n x_i * y_i * (u_i^2 + v_i^2) \right) - T_{11}^* \left( \sum_{i=1}^n x_i * z_i * (u_i^2 + v_i^2) \right) = \sum_{i=1}^n x_i * (u_i^2 + v_i^2) \quad (\text{E1})
\end{aligned}$$

$$\begin{aligned}
& T_1^* \left( \sum_{i=1}^n x_i * y_i * u_i \right) + T_2^* \left( \sum_{i=1}^n y_i^2 * u_i \right) + T_3^* \left( \sum_{i=1}^n y_i * z_i * u_i \right) + T_4^* \left( \sum_{i=1}^n y_i * u_i \right) + \\
& + T_5^* \left( \sum_{i=1}^n x_i * y_i * v_i \right) + T_6^* \left( \sum_{i=1}^n y_i^2 * v_i \right) + T_7^* \left( \sum_{i=1}^n y_i * z_i * v_i \right) + T_8^* \left( \sum_{i=1}^n y_i * v_i \right) - \\
& - T_9^* \left( \sum_{i=1}^n x_i * y_i * (u_i^2 + v_i^2) \right) - T_{10}^* \left( \sum_{i=1}^n y_i^2 * (u_i^2 + v_i^2) \right) - T_{11}^* \left( \sum_{i=1}^n y_i * z_i * (u_i^2 + v_i^2) \right) = \sum_{i=1}^n y_i * (u_i^2 + v_i^2) \quad (\text{E2})
\end{aligned}$$

$$\begin{aligned}
& T_1^* \left( \sum_{i=1}^n x_i * z_i * u_i \right) + T_2^* \left( \sum_{i=1}^n y_i * z_i * u_i \right) + T_3^* \left( \sum_{i=1}^n z_i^2 * u_i \right) + T_4^* \left( \sum_{i=1}^n z_i * u_i \right) + \\
& + T_5^* \left( \sum_{i=1}^n x_i * z_i * v_i \right) + T_6^* \left( \sum_{i=1}^n y_i * z_i * v_i \right) + T_7^* \left( \sum_{i=1}^n z_i^2 * v_i \right) + T_8^* \left( \sum_{i=1}^n z_i * v_i \right) - \\
& - T_9^* \left( \sum_{i=1}^n x_i * z_i * (u_i^2 + v_i^2) \right) - T_{10}^* \left( \sum_{i=1}^n y_i * z_i * (u_i^2 + v_i^2) \right) - T_{11}^* \left( \sum_{i=1}^n z_i^2 * (u_i^2 + v_i^2) \right) = \sum_{i=1}^n z_i * (u_i^2 + v_i^2) \quad (\text{E3})
\end{aligned}$$

$$T_1^* \left( \sum_{i=1}^n x_i^2 \right) + T_2^* \left( \sum_{i=1}^n x_i * y_i \right) + T_3^* \left( \sum_{i=1}^n x_i * z_i \right) + T_4^* \left( \sum_{i=1}^n x_i \right) -$$

$$-T_9^* \left( \sum_{i=1}^n x_i^2 * u_i \right) - T_{10}^* \left( \sum_{i=1}^n x_i * y_i * u_i \right) - T_{11}^* \left( \sum_{i=1}^n x_i * z_i * u_i \right) = \sum_{i=1}^n (x_i * u_i) \quad (E4)$$

$$T_1^* \left( \sum_{i=1}^n x_i * y_i \right) + T_2^* \left( \sum_{i=1}^n y_i^2 \right) + T_3^* \left( \sum_{i=1}^n y_i * z_i \right) + T_4^* \left( \sum_{i=1}^n y_i \right) -$$

$$-T_9^* \left( \sum_{i=1}^n x_i * y_i * u_i \right) + T_{10}^* \left( \sum_{i=1}^n y_i^2 * u_i \right) + T_{11}^* \left( \sum_{i=1}^n y_i * z_i * u_i \right) = \sum_{i=1}^n (y_i * u_i) \quad (E5)$$

$$T_1^* \left( \sum_{i=1}^n x_i * z_i \right) + T_2^* \left( \sum_{i=1}^n y_i * z_i \right) + T_3^* \left( \sum_{i=1}^n z_i^2 \right) + T_4^* \left( \sum_{i=1}^n z_i \right) -$$

$$-T_9^* \left( \sum_{i=1}^n x_i * z_i * u_i \right) - T_{10}^* \left( \sum_{i=1}^n y_i * z_i * u_i \right) - T_{12}^* \left( \sum_{i=1}^n z_i^2 * u_i \right) = \sum_{i=1}^n (z_i * u_i) \quad (E6)$$

$$T_1^* \left( \sum_{i=1}^n x_i \right) + T_2^* \left( \sum_{i=1}^n y_i \right) + T_3^* \left( \sum_{i=1}^n z_i \right) + T_4^* * n -$$

$$-T_9^* \left( \sum_{i=1}^n x_i * u_i \right) - T_{10}^* \left( \sum_{i=1}^n y_i * u_i \right) - T_{11}^* \left( \sum_{i=1}^n z_i * u_i \right) = \sum_{i=1}^n u_i \quad (E7)$$

$$T_5^* \left( \sum_{i=1}^n x_i^2 \right) + T_6^* \left( \sum_{i=1}^n x_i * y_i \right) + T_7^* \left( \sum_{i=1}^n x_i * z_i \right) + T_8^* \left( \sum_{i=1}^n x_i \right) -$$

$$-T_9^* \left( \sum_{i=1}^n x_i^2 * v_i \right) - T_{10}^* \left( \sum_{i=1}^n x_i * y_i * v_i \right) - T_{11}^* \left( \sum_{i=1}^n x_i * z_i * v_i \right) = \sum_{i=1}^n (x_i * v_i) \quad (E8)$$

$$T_5^* \left( \sum_{i=1}^n x_i * y_i \right) + T_6^* \left( \sum_{i=1}^n y_i^2 \right) + T_7^* \left( \sum_{i=1}^n y_i * z_i \right) + T_8^* \left( \sum_{i=1}^n y_i \right) -$$

$$-T_9^* \left( \sum_{i=1}^n x_i * y_i * v_i \right) + T_{10}^* \left( \sum_{i=1}^n y_i^2 * v_i \right) + T_{11}^* \left( \sum_{i=1}^n y_i * z_i * v_i \right) = \sum_{i=1}^n (y_i * v_i) \quad (E9)$$

$$\begin{aligned}
& T_5^* \left( \sum_{i=1}^n x_i^* z_i \right) + T_6^* \left( \sum_{i=1}^n y_i^* z_i \right) + T_7^* \left( \sum_{i=1}^n z_i^2 \right) + T_8^* \left( \sum_{i=1}^n z_i \right) - \\
& - T_9^* \left( \sum_{i=1}^n x_i^* z_i^* v_i \right) - T_{10}^* \left( \sum_{i=1}^n y_i^* z_i^* v_i \right) - T_{12}^* \left( \sum_{i=1}^n z_i^2^* v_i \right) = \sum_{i=1}^n (z_i^* v_i) \quad (\text{E10})
\end{aligned}$$

$$\begin{aligned}
& T_5^* \left( \sum_{i=1}^n x_i \right) + T_6^* \left( \sum_{i=1}^n y_i \right) + T_7^* \left( \sum_{i=1}^n z_i \right) + T_8^* n - \\
& - T_9^* \left( \sum_{i=1}^n x_i^* v_i \right) - T_{10}^* \left( \sum_{i=1}^n y_i^* v_i \right) - T_{11}^* \left( \sum_{i=1}^n z_i^* v_i \right) = \sum_{i=1}^n v_i \quad (\text{E11})
\end{aligned}$$

#### 4. Estimating the "quality" of the results

We would like to lump somehow the non-linear nature ( cf. equations ( 17 ) ) of the original problem in a single parameter. At this point one should also recall the error

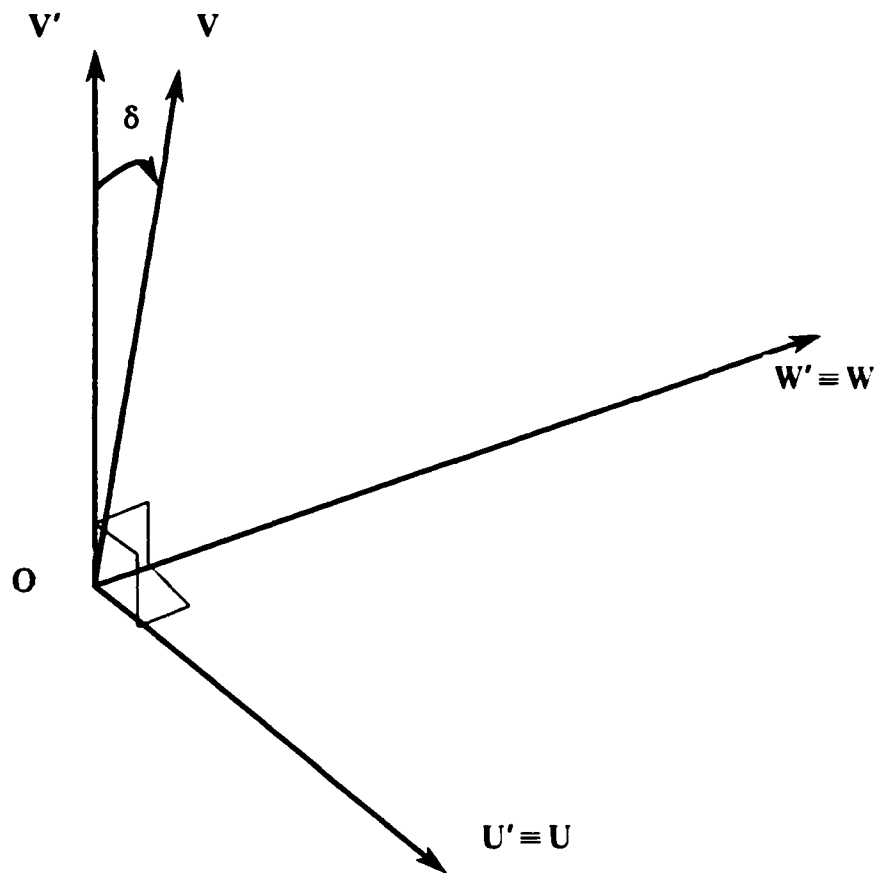


Fig. 2

introduced by the observations ( i.e.  $(u_i, v_i)$  &  $(x_i, y_i, z_i)$  ) made as well as the fact that the camera used is not a pin hole one without distortions. Assume that all these errors that are intrinsic to the experimental approach amount to the determination, using equations ( 47 ) - ( 54 ), to a coordinate system UVW such that axes U and V are not perpendicular to each other but slightly off by a small angle  $\delta$  ( see the above Fig. 2 ). We furthermore considered axes U'V'W', that constitute an orthonormal

coordinate system, such that the pairs ( W , W' ) and (V, V' ) of axes represent axes parallel to each other. It is clear that the coordinates u,v,w can still be expressed in terms of x,y,z ( see also Fig. 1 ) as follows:

$$\begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} = \begin{bmatrix} a' & b' & c' \\ d' & e' & f' \\ g' & h' & i' \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} \quad (71)$$

where a', b', c', d', e', f', g', h', i' are the coefficients of a proper rotation matrix. Moreover we observe that we can write:

$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ \sin\delta & \cos\delta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} \quad (72)$$

Hence we can write the following:

$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} a' & b' & c' \\ a' * \sin\delta + d' * \cos\delta & b' * \sin\delta + e' * \cos\delta & c' * \sin\delta + f' * \cos\delta \\ g' & h' & i' \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} \quad (73)$$

Comparing now equations (72) and ( 73 ) we observe that we can write:

$$a = a'$$

$$b = b'$$

$$c = c'$$

$$d = a' \sin \delta + d' \cos \delta$$

$$e = b' \sin \delta + e' \cos \delta$$

$$f = c' \sin \delta + f' \cos \delta$$

$$g = g'$$

$$h = h'$$

$$i = i'$$

Therefore,

$$a * d + b * e + c * f = \sin \delta * ( a'^2 + b'^2 + c'^2 ) + \cos \delta * ( a' * d' + b' * e' + c' * f' ) = \sin \delta$$

Recall that from equation ( 23 ) - "normality" equation - we have :

$$a * d + b * e + c * f = 0$$

Hence, we can estimate the " goodness " of the calibration using the value of parameter  $\delta$ . It has been observed, that the closer  $\delta$ 's value is to 0 the better the calibration under consideration. In fact, values of  $\delta \sim 10^{-3}$  radians are considered satisfactory for solid state as well as vidicon cameras. In case  $\delta$  is significantly greater than  $10^{-3}$ , it is better to consider a new set of points than to attempt to recalibrate the camera with the same set of points.

## 5. Deriving more important parameters of the problem

So far we have solved the linear system that resulted after the consideration of the initial equations of the problem. In other words we have computed the entries of the matrix  $M_3$ , or equivalently the unknowns  $T_i$ 's /  $i=1,2,3 \dots 12$ , ( recall that  $T_{12}=1$  ). We still need to translate these coefficients into camera parameters, namely the position of the center of the camera in 3-D, the scale factors  $k_u$  and  $k_v$  as well as the pair  $(u_0, v_0)$  ( see page 6 ).

The  $u_0$  and  $v_0$  are computed using equations ( 44 ) and ( 45 ). Recalling the definitions of  $k_u$  and  $k_v$  we observe that using the results of equations ( 45 ) and ( 46 ) we can compute  $k_u$  and  $k_v$  as follows:

$$k_u = k_1 / fl$$

$$k_v = k_2 / fl$$

where  $fl$  is the ( known ) focal length of the camera.

Furthermore, in what concerns the position of the camera in 3-D, one need only recall equations ( 8 ), ( 9 ), and ( 10 ). These equations directly give us the 3-D coordinates of the center of the camera.

At this point it is worth recalling the discussion about the sign of  $k_2$ . As we saw, according to the set of points considered, the sign of  $k_2$  might be different each time. And since the values of  $d$ ,  $e$ ,  $f$ , and  $q$  all depend on the values of  $k_2$  it seems that we must determine each time the sign of  $k_2$ , in order to compute  $x_0$ ,  $y_0$ ,  $z_0$ . However we can observe that all these variables appear in equations ( 8 ), ( 9 ), ( 10 ), as products of pairs and therefore their signs always reduce to " + ". Hence there is no need at all to compute the sign of  $k_2$ .

Finally, we could also determine the directions of the axes  $U$ ,  $V$ , and  $W$  with respect to the fixed 3-D coordinate system  $OXYZ$ . In order to do this we only need to consider the ( orthonormal ) vectors:

$$n_1 = ( a, b, c )$$

$$n_2 = ( d, e, f )$$

$$\mathbf{n}_3 = (g, h, i)$$

As a matter of fact  $\mathbf{n}_1$  is the vector that is parallel to U-axis,  $\mathbf{n}_2$  is the vector that is parallel to the V-axis and  $\mathbf{n}_3$  is the vector that is parallel to the W-axis.

*Note: As it has been also pointed out by Ganapathy ([5]), the computation of  $u_0, v_0$  is rather sensitive. Therefore, if one is interested in a quite precise computation of those quantities, one should follow another approach ( see APPENDIX I ).*

## 6. Deficiencies of the algorithm

The above described algorithm has one important deficiency. Assume that the considered set of points has a symmetry with respect to any of the planes defined by any two axes of the 3-D coordinate system. If the optical axis lies on this plane, then some of the summations appearing in equations ( E1 ) through ( E11 ) will be equal to 0. Which ones exactly are equal to 0, depends on the symmetry. However, this is enough to reduce the rank of the coefficients' matrix ( matrix  $A_{ij}$  below ). But reduction of the rank of this matrix has as a consequence our inability to solve the linear system of the 11 equations in the 11 unknowns  $T_i, i=1, 2, 3, .. 11$ . It is therefore clear that one should not choose such configurations of points. However,

$$\begin{bmatrix} A_{11} & A_{12} & \dots & A_{1m} \\ A_{21} & A_{22} & \dots & A_{2m} \\ \cdot & \cdot & & \cdot \\ \cdot & \cdot & & \cdot \\ \cdot & & & \cdot \\ \cdot & & & \cdot \\ A_{n1} & & & A_{nm} \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ \cdot \\ \cdot \\ \cdot \\ X_m \end{bmatrix} = \begin{bmatrix} C_1 \\ C_2 \\ C_3 \\ \cdot \\ \cdot \\ \cdot \\ C_m \end{bmatrix}$$

the problem is not necessarily solved. This fact lies on more intrinsic reasons. Let us consider the above linear system of equations.

*Definition:* The above system of equations is called **ill-conditioned** if the determinant of the matrix of coefficients has a value that is very close to 0.

In such cases, an error is introduced in the values of the computed parameters. Moreover, consider the case where the coefficients  $A_{ij}$  are measured with an error  $\pm \Delta A_{ij}$ . This means that the value of  $A_{ij}$  lies in an interval  $[A^*_{ij} - \epsilon_{ij}, A^*_{ij} + \epsilon_{ij}]$ ,  $i = 1, 2, 3, \dots, n$  &  $j = 1, 2, 3, \dots, m$ . The "stared" values  $A^*_{ij}$  are the ones that "correct" measurements should give. It is clear that we cannot be sure that for *no set of values*  $A_{ij}$ , the above system become ill-conditioned.

*Definition:* We call a linear system **critically ill-conditioned** if it is not ill-conditioned, but there is at least one set of values for  $A_{ij}$  in the intervals  $[A^*_{ij} - \epsilon_{ij}, A^*_{ij} + \epsilon_{ij}]$ ,  $i = 1, 2, 3, \dots, n$  &  $j = 1, 2, 3, \dots, m$ , for which the linear system becomes ill-conditioned.

In a recent work ([1]), that dealt with this problem, a measure of ill-conditioning was determined. In fact, the quantity:

$$\phi = \sum_{i=1}^n \sum_{j=1}^m |b_{ij}| \cdot \epsilon_{ij}$$

where  $b_{ij} = ((A^{-1})^T)_{ij}$   $i = 1, 2, 3, \dots, n$  &  $j = 1, 2, 3, \dots, m$ , can be used to determine whether the considered system is ill-conditioned. The following hold:

$$\begin{aligned} \phi < 1 &\Rightarrow \text{Not critically ill-conditioned} \\ \phi \geq 1 &\Rightarrow \text{Critically ill-conditioned} \end{aligned}$$

Furthermore, an upper bound for the error  $\Delta X_i$   $i = 1, 2, 3, \dots, m$ , in the computed values of  $X_k$   $k = 1, 2, 3, \dots, m$ , was established:

$$\Delta X_i \leq \frac{\phi}{1 - \phi} \|x\|_1 \quad i = 1, 2, 3, \dots, m$$

where  $\|a\|_1$  is the first norm of matrix  $a = (a_{jk}) / j=1,2,3, \dots n \ \& \ k=1,2,3, \dots m$ , that is defined as follows:

$$\|a\|_1 \equiv \max_k \left\{ \sum_{j=1}^n |a_{jk}| \right\} \text{ or } \max_j \left\{ \sum_{k=1}^m |a_{jk}| \right\}$$

One could use the parameter  $\phi$ , to determine whether the system of equations (E1) thru (E11), is ill-conditioned or not, and therefore determine the "quality" of the calibration. However, the necessary overhead does not make the effort worth it, since one can easily tell from the calibration matrix whether the system is ill-conditioned (in which case most of its entries will be equal to 0), or not.

## 7. Implementation of the described algorithm

The algorithm we just described, has been implemented in the University of Rochester, during the spring of 1986. A full description of the program's output together with useful hints and other relevant information can be found in APPENDIX II.

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## APPENDIX I

The following describes a quick method for the determination of the coordinates of the pixel on the DataCube display through which the optical axis of the camera in use passes. The idea of the method relies on the notion and the properties of the so-called "vanishing point".

**Method:** A set of 8 points forming the vertices of a cube is constructed in 3-D space (see Fig. 1.a ). These points can be constructed using 8 nuts screwed on vertical threaded rods, or using a set of 8 targets like the one shown in Fig. 1.b. The camera under consideration is moved until the points A, A', B, B', C, C', D, D' of the image plane ( Datacube display ) that correspond to the "vertices" of the original cube, are such, that the quadruples ( A, A', C, C' ) and ( B, B', D, D' ) represent sets of points lying on two distinct lines of the image, say  $\epsilon_1$  and  $\epsilon_2$  respectively ( see Fig. 2.a and Fig. 2.b ). Lines  $\epsilon_1$  and  $\epsilon_2$  intersect at a point P ( the vanishing point ). The coordinates of this point are the coordinates we are after. How can one determine the coordinates of P? Recall that A, A', C, C' lie on the same line, namely  $\epsilon_1$ . Using a cursor moving routine, one determines the coordinates of points A, A', C, C'. Then using a simple calculator, and applying a least-squares fitting method, the slope and the offset of line  $\epsilon_1$  are computed. The same procedure is repeated for the case of line  $\epsilon_2$ . The slopes and offsets of lines  $\epsilon_1$  and  $\epsilon_2$  being determined, the coordinates of P can be computed; in fact one need only express  $\epsilon_1$  and  $\epsilon_2$  as linear equations in two unknowns, say x and y, and then solve the resulting linear system. The solution of this system will be the pair of coordinates of P.

Improvement of the results can be achieved by repeating the above procedure using other rigid configurations ( see Fig. 3.a and Fig. 3.b ).

*Note: Obviously, the above described method is general, and can be applied to any camera and display combination.*

**Results:** Application of the above described method for the case of a cube gave the following results, for the cameras of our laboratory:

<u>CAMERA TYPE</u>	<u>COORDINATES</u>
vidicon	( 353, 247 )
ccd	( 334, 200 )

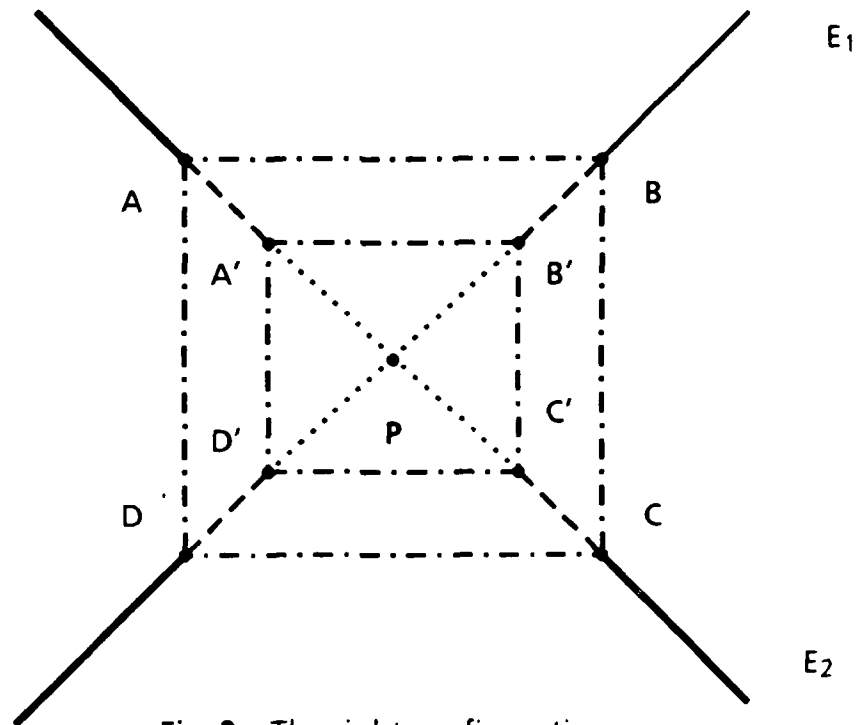


Fig. 2.a The right configuration

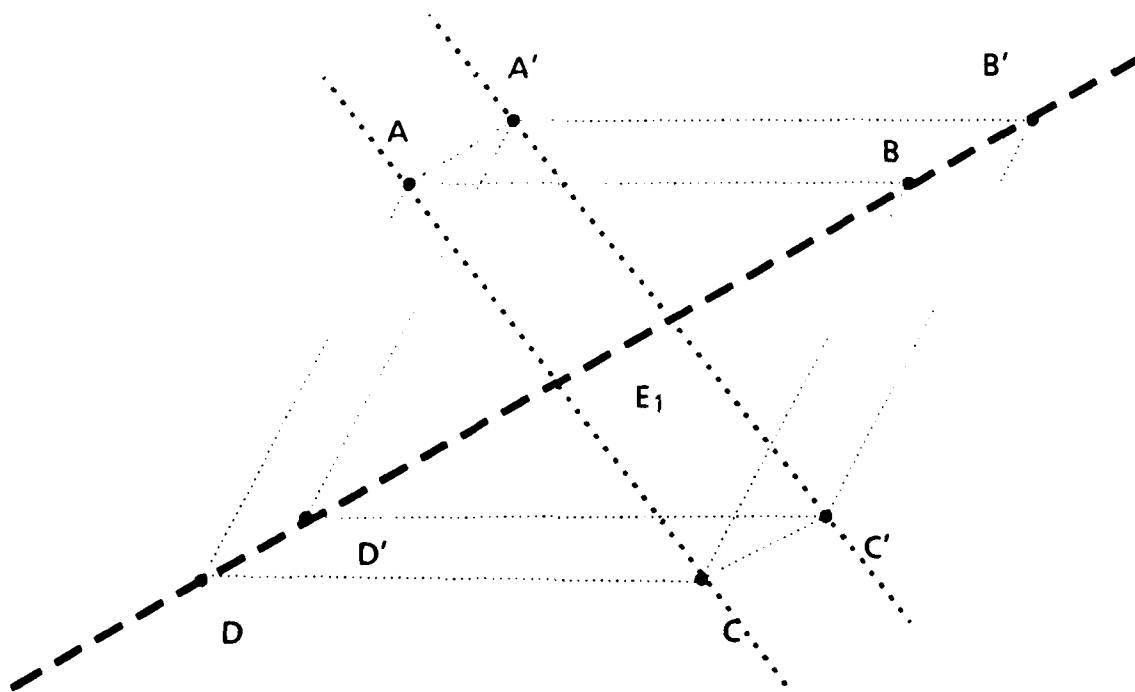


Fig. 2.b A wrong configuration.

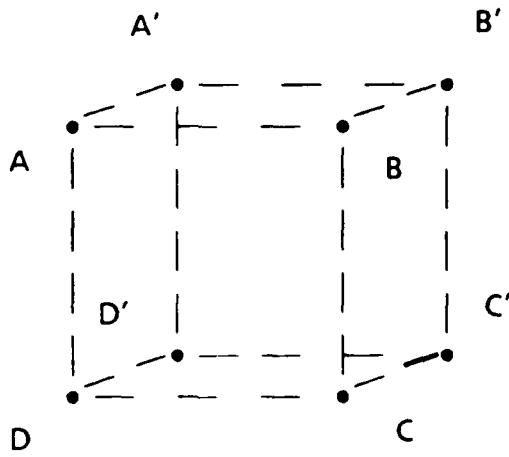


Fig. 1.a

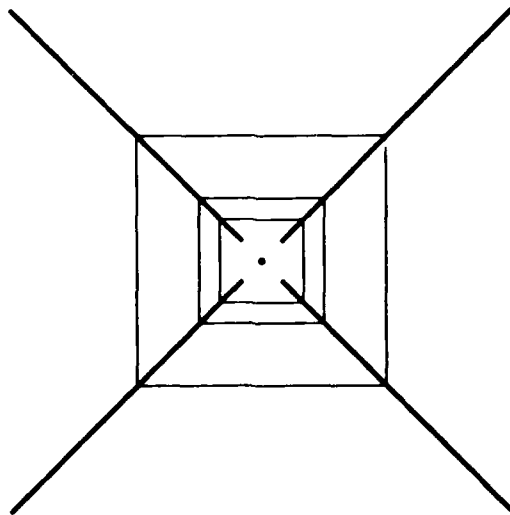


Fig. 1.b

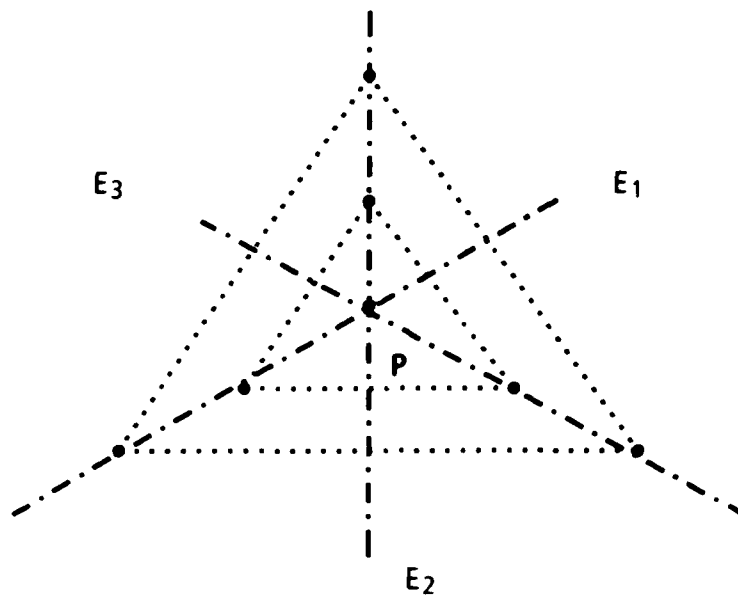


Fig. 3.a Case of a prism.

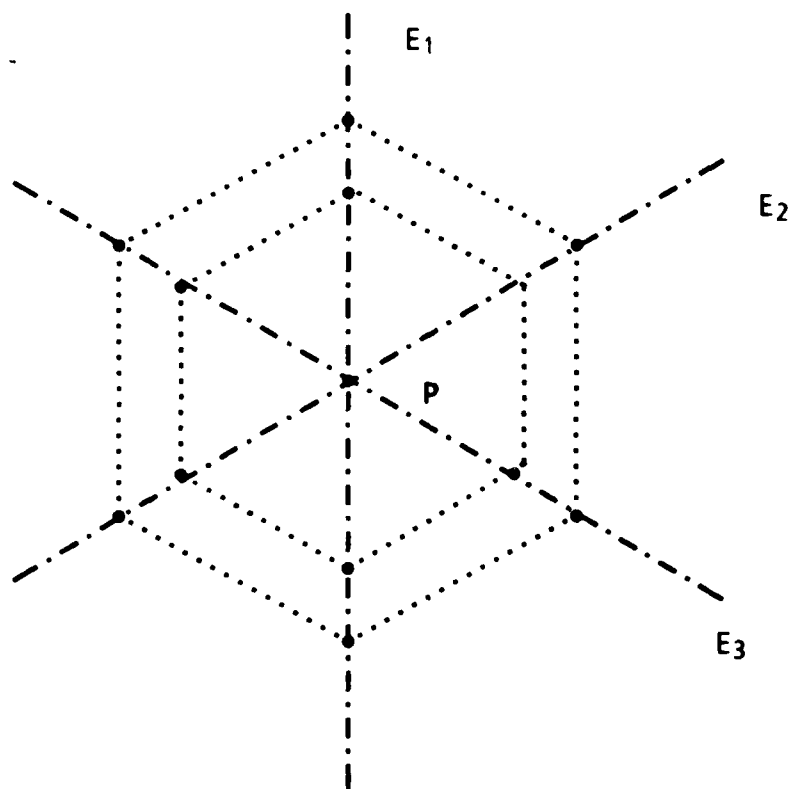


Fig. 3.b Case of an octahedron.

## APPENDIX II

This appendix gives a) some very useful observations that have been made as well as hints, and b) the file structures that are needed together with a description of the output of the actual program.

### Part I

#### Choosing the 3-D points

This is the most crucial part of the whole procedure. In fact the selection of the world points can lead to a disaster or a success. Some useful observations have been made:

- a) choose a sufficient number of points; although the algorithm requires only 5 <sup>1</sup>/<sub>2</sub> points, a "trustworthy" calibration was found to require at least 8 or 9 points.
- b) pay attention to the configuration of the 3-D points with respect to the optical axis and the 3-D ( fixed ) coordinate system; besides some obvious degenerate cases for which the algorithm will not work, there are also other cases that may cause problems. As a general suggestion, though, one should keep in mind that the more randomly the targets are selected, the higher the possibility for a succesful calibration. It is obvious, of course that one should not select configurations that experience some kind of symmetry ( especially axial-symmetry ) with respect to the optical axis.
- c) the 3-D points should be selected "*as far as possible*" from the camera under consideration; of course this is in direct relation with the focal length of the camera. In general, the targets must be placed at such positions, so that all of them are focused. Obviously, if the camera supports apperture setting facilities, one would like to set the apperture to 16 or 22 under intense lighting conditions.

#### Measuring coordinates

① 3-D space: It has been observed that measuring the 3-D coordinates of the world points with an accuracy of approximately 0.1 inches is sufficiently satisfactory. However, one should keep in mind that all the quantities that have a dimension of *MOLITO*, e.g. focal length of the camera, world-point coordinates etc., must be expressed in the same length units, e.g. all in inches or all in millimeters ...

② Image plane: Using a cursor moving routine, one must determine the image plane coordinates with an inaccuracy of at most  $\pm 3$  pixels.

### Discovering potential deficiencies of the camera in use

One problem that might arise is the following: "cheap optics" or "not properly mounted on the camera" lens. One could easily figure out whether this is a problem for the camera in use, or not, by placing some targets in front of the camera and trying to focus on them; if their images are displaced in a **non-consistent** way, then there is a high possibility of the camera having the above mentioned problem. Of course it is needless to say that there is no hope of calibrating this camera .

### Algorithm's results

It has been observed that the computation of the values  $u_0$ ,  $v_0$ , is rather sensible and related to the direction of the optical axis. In fact, one should use another approach in order to get more robust values for these parameters ( see APPENDIX I). Besides that, the remaining parameters are precisely computed. It must be pointed out, though, that if one wants to get reliable results, one must try a few calibrations with the camera being horizontal, and for different camera positions ( with respect to the 3-D coordinate frame ) and sets of world-points, in order to get a feeling about the the response of the camera in use. Then other positions might be tried ( pan, tilt or roll  $\neq 0$  ) as well.

## Part II

### Description of the necessary file structures

The files that are used by the calibration routine are two:

i) the file " data3D ", that contains the number of experimental points as well as their 3D coordiates. This file must be set up by the user, and should have the following format:

```
line #1          #of__points__used
line #2   X1 ( space ) Y1 ( space ) Z1
line #3   X2 ( space ) Y2 ( space ) Z2
line #4   X3 ( space ) Y3 ( space ) Z3
```

etc.

*Note: The X<sub>i</sub>'s, Y<sub>i</sub>'s and Z<sub>i</sub>'s are read as " double's ".*

ii) the file " dataimage " that contains the image plane coordinates of the experimental points. This file is either set up by the user, in which case it has to have the following format:

```
line #1    u1 ( space ) v1  
line #2    u2 ( space ) v2  
line #3    u3 ( space ) v3  
line #4    u4 ( space ) v4
```

etc.

or by invoking the special command called " dqstorepoint "; the latter is a cursor moving routine that uses the "mouse" of a SUN-Workstation that is connected to the DataCube. The user:

- 1) selects points by clicking the left button of the mouse
- 2) confirms a selected point and makes an entry for the corresponding coordinates to the file " dataimage ", by clicking the middle button of the "mouse"
- 3) " exits " by clicking the right button of the "mouse".

*Note: It is worth mentioning that the entries of the ( i + 1 )-th line of the file " data3D " and the ones of the i-th line of the file " dataimage " correspond to the i-th experimental point, i = 1,2,3,4,5,6, ... n.*

### A typical session

The following is a script of a typical session. We assume that the files " data3D " and " dataimage " were appropriately set up as described above. The names of the parameters are consistent with the ones appearing to the description of the algorithm...

```
< ISIDORE >>> % cat data3D  
9  
552.2 376.5 1989.1  
815.75 376.5 1989.1  
815.75 169.4 1989.1  
351.4 169.4 1689.3  
351.4 376.5 1689.3  
702.8 169.4 1185.975
```

702.8 376.5 1185.975  
753.0 169.4 1085.575  
753.0 376.5 1085.575  
< ISIDORE >>> % cat dataimage  
286 164  
504 164  
505 318  
62 327  
64 143  
432 372  
429 89  
517 385  
517 67  
< ISIDORE >>> % dqcalibrate

Please enter the focal length of the camera in use: 25

Reading the 3-D coordinates Number of points= 9

Point number 1  
552.200000 376.500000 1989.100000

Point number 2  
815.750000 376.500000 1989.100000

Point number 3  
815.750000 169.400000 1989.100000

Point number 4  
351.400000 169.400000 1689.300000

Point number 5  
351.400000 376.500000 1689.300000

Point number 6  
702.800000 169.400000 1185.975000

Point number 7  
702.800000 376.500000 1185.975000

Point number 8  
753.000000 169.400000 1085.575000

Point number 9  
753.000000 376.500000 1085.575000

Reading the image-plane coordinates

Point number 1  
286.000000 164.000000

Point number 2  
504.000000 164.000000

Point number 3  
505.000000 318.000000

Point number 4  
62.000000 327.000000

Point number 5  
64.000000 143.000000

Point number 6  
432.000000 372.000000

Point number 7

429.000000 89.000000

Point number 8

517.000000 385.000000

Point number 9

517.000000 67.000000

The solution matrix is the following :

-5.288670	-0.060201	-1.428586	3957.598778
-0.051500	4.650256	-0.929366	-914.765229
-0.000063	-0.000189	-0.003655	1.000000

r2 = 74628.615332

Observe  $ad + be + cf = 0.011060$

The skew angle estimation parameter is equal (in radians) to  
: 0.011060

u0=415.503321 v0=188.044822

The resolution parameters are:

Ku=57.512410 Kv=51.273858

The remaining parameters are :

a=-0.999847 b=0.003509 c=0.017124

d=-0.008440 e=0.998633 f=-0.051583

g=-0.017282 h=-0.051723 i=-0.998512

p=672.994285 q=-235.026810 r=273.182385

The directions of the 3 axes with respect to the fixed 3D coordinate system are

U-axis: (-0.999847, 0.003509, 0.017124)

V-axis: (-0.008440, 0.998633, -0.051583)

(optical)Z-axis: (-0.017282, -0.051723, -0.998512)

The position of the center of the camera is at:

Xo=675.628994 Yo=246.473854 Zo=249.128475

< ISIDORE >>> %

END

DATE

3-88

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