

Attitude Determination in Higher Dimensions

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Introduction

GENERALLY, one has little cause to estimate an attitude in spaces of dimension higher than three. This exercise, however, will afford an insight into the workings of a well-known attitude determination algorithm in three dimensions. In addition, should the dimensionality of our world ever increase without notice, we will be all the better prepared.

An $n \times n$ proper orthogonal matrix A satisfies

$$A^T A = I_{n \times n} \quad (1)$$

$$\det A = 1 \quad (2)$$

Equation (1) is equivalent to $n(n+1)/2$ constraints on the matrix A . Hence, A can have only $n(n-1)/2$ free parameters, as remarked by Bar-Itzhack¹ and Bar-Itzhack and Markley.² Thus, A may be represented in terms of matrices of manifestly smaller parameter dimension. For example,

$$A = \exp\{\Theta\} \quad (3)$$

where Θ is an $n \times n$ antisymmetric matrix whose independent elements are the n -dimensional generalization of the rotation vector.¹ Likewise, one may write²

$$A = (I + G)(I - G)^{-1} \quad (4)$$

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where G is an $n \times n$ antisymmetric matrix, whose independent elements are the n -dimensional equivalent of the Gibbs vector.^{3,4} The matrices Θ and G are related by

$$G = \tanh(\Theta/2) \quad (5)$$

Writing

$$G = \{[g]\}, \quad \Theta = \{[\theta]\} = \{[\theta\hat{n}]\} \quad (6)$$

where

$$\{[u]\} = \begin{bmatrix} 0 & u_3 & -u_2 \\ -u_3 & 0 & u_1 \\ u_2 & -u_1 & 0 \end{bmatrix} \quad (7)$$

the equations in n dimensions assume their familiar three-dimensional forms,⁴

$$A = \exp(\{[\theta]\}) \quad (8)$$

$$A = (I + \{[g]\})(I - \{[g]\})^{-1} \quad (9)$$

and

$$g = \tan(\theta/2)\hat{n} \quad (10)$$

(Note that we have chosen a different sign convention than Bar-Itzhack and Markley.)

Attitude Determination Problem

Suppose that we are given N linearly independent vector measurements W_k , $k = 1, \dots, N$, which are the representations of N n -dimensional vectors in the spacecraft reference frame. These are related to V_k , $k = 1, \dots, N$, the representations of these same vectors in the primary reference frame according to

$$W_k = AV_k, \quad k = 1, \dots, N \quad (11)$$

where A is assumed to be an $n \times n$ proper orthogonal matrix. Equation (11) assumes that the measurements are perfect, i.e., noise free, which should be no more true in practice in n dimensions than it is in three dimensions. (Fortunately, there is not a well-established practice for dimensions higher than three.) The assumption implied by Eq. (11), however, is similar to the one behind the triad method,^{5,6} so that Eq. (11) is not without precedent.

The first question to be answered is, What is the minimum value of N that permits a unique solution? We remark first that the N vectors amount to (Nn) total components. At the same time, from Eq. (11), the N vectors are subject to $N(N+1)/2$ constraints of the form

$$W_i \cdot W_j = V_i \cdot V_j, \quad i = 1, \dots, N, \quad j = 1, \dots, N \quad (12)$$

Thus, the total number of unconstrained components in the vector measurements, i.e., the number of components that carry independent information about the attitude, is $Nn - N(N+1)/2$. The minimum number of measurements, then, is the minimum value of N satisfying

$$Nn - \frac{N(N+1)}{2} \geq \frac{n(n-1)}{2} \quad (13)$$

The equality, in fact, has two solutions, $n-1$ and n . Thus,

$$N_{\min} = n - 1 \quad (14)$$

Ebert⁷ has offered a geometrical argument for this result. Suppose that we are given $n-m$ vector measurements. These $n-m$ vectors span a subspace \mathcal{V}_{n-m} of dimension $n-m$ of the

n -dimensional vector space \mathcal{V}_n . The vector space \mathcal{V}_n may be written, therefore, as the direct sum $\mathcal{V}_n = \mathcal{V}_{n-m} \oplus \mathcal{V}_m$ of this subspace and the subspace of vectors perpendicular to the $n-m$ measurements. Clearly, any rotation within \mathcal{V}_m will leave the $n-m$ measurements unchanged. Thus, for the $n-m$ measured vectors to define the attitude uniquely, it is necessary that the dimension of \mathcal{V}_m be so small that rotations are not possible. Thus, we must have either $m=0$ or 1 . The latter value leads to the smaller value of $n-m$ and Eq. (14).

Thus, the attitude determination problem is to determine an $n \times n$ proper orthogonal matrix A given $n-1$ linearly independent n -dimensional vector measurements W_k , $k = 1, \dots, n-1$.

General n -Dimensional Algorithm

Given the $n-1$ vectors V_k , $k = 1, \dots, n-1$, we construct a set of $n-1$ orthonormal vectors \hat{r}_k , $k = 1, \dots, n-1$, according to the Gram-Schmidt orthogonalization.⁸ Thus,

$$\begin{aligned} r_1 &= V_1, & \hat{r}_1 &= r_1/|r_1| \\ r_2 &= V_2 - (\hat{r}_1 \cdot V_2)\hat{r}_1, & \hat{r}_2 &= r_2/|r_2| \\ & \vdots & & \vdots \\ r_{n-1} &= V_{n-1} - \sum_{k=1}^{n-2} (\hat{r}_k \cdot V_{n-1})\hat{r}_k, & \hat{r}_{n-1} &= r_{n-1}/|r_{n-1}| \end{aligned} \quad (15)$$

Similarly, we construct $n-1$ orthonormal vectors \hat{s}_k , $k = 1, \dots, n-1$, from the W_k , $k = 1, \dots, n-1$. It follows from Eq. (11) that

$$\hat{s}_k = A\hat{r}_k \quad (16)$$

We now construct the vector \hat{r}_n (and in a corresponding fashion \hat{s}_n) according to

$$(\hat{r}_n)_\ell = \sum_{i_1, i_2, \dots, i_{n-1}} \epsilon_{i_1 i_2 \dots i_{n-1} \ell} (\hat{r}_1)_{i_1} (\hat{r}_2)_{i_2} \dots (\hat{r}_{n-1})_{i_{n-1}} \quad (17)$$

where $(\hat{r}_k)_m$ denotes the m th component of \hat{r}_k , and the sums over each index are from 1 to n . The quantity $\epsilon_{i_1 i_2 \dots i_n}$ is the Levi-Civita symbol in n indices, which is just the parity of the permutation taking $(1, 2, \dots, n)$ into (i_1, i_2, \dots, i_n) and vanishes if the latter is not a permutation of the former. Thus, the Levi-Civita symbol satisfies

$$\epsilon_{123 \dots n} = 1 \quad (18)$$

$$\epsilon_{i_1 \dots i_j i_{j+1} \dots i_n} = -\epsilon_{i_1 \dots i_{j+1} i_j \dots i_n} \quad (19)$$

The determinant of an $n \times n$ matrix M is often defined in terms of the Levi-Civita symbol⁹ as

$$\det M = \sum_{i_1, i_2, \dots, i_n} \epsilon_{i_1 i_2 \dots i_n} M_{1i_1} M_{2i_2} \dots M_{ni_n} \quad (20)$$

where M_{ki} denotes the (k, i) element of M . From Eq. (17)

$$\begin{aligned} (\hat{s}_n)_\ell &= \sum_{i_1, i_2, \dots, i_{n-1}} \epsilon_{i_1 i_2 \dots i_{n-1} \ell} (A\hat{r}_1)_{i_1} (A\hat{r}_2)_{i_2} \dots (A\hat{r}_{n-1})_{i_{n-1}} \\ &= \sum_{i_1, i_2, \dots, i_{n-1}} \epsilon_{i_1 i_2 \dots i_{n-1} \ell} \sum_{j_1, j_2, \dots, j_{n-1}} A_{i_1 j_1} A_{i_2 j_2} \dots A_{i_{n-1} j_{n-1}} \\ &\quad \times (\hat{r}_1)_{j_1} (\hat{r}_2)_{j_2} \dots (\hat{r}_{n-1})_{j_{n-1}} \end{aligned} \quad (21)$$

Thus,

$$\begin{aligned} (A^T \hat{s}_n)_\ell &= \sum_{i_n} (A^T)_{i_n \ell} (\hat{s}_n)_{i_n} = \sum_{i_n} A_{i_n \ell} (\hat{s}_n)_{i_n} = \sum_{i_1, i_2, \dots, i_{n-1}} \epsilon_{i_1 i_2 \dots i_{n-1} \ell} \\ &\quad \times \sum_{j_1, j_2, \dots, j_{n-1}} A_{i_1 j_1} A_{i_2 j_2} \dots A_{i_{n-1} j_{n-1}} A_{i_n \ell} \\ &\quad \times (\hat{r}_1)_{j_1} (\hat{r}_2)_{j_2} \dots (\hat{r}_{n-1})_{j_{n-1}} = (\det A) \sum_{j_1, j_2, \dots, j_{n-1}} \epsilon_{j_1 j_2 \dots j_{n-1} \ell} \\ &\quad \times (\hat{r}_1)_{j_1} (\hat{r}_2)_{j_2} \dots (\hat{r}_{n-1})_{j_{n-1}} = (\hat{r}_n)_\ell \end{aligned} \quad (22)$$

so that

$$\hat{s}_n = A\hat{r}_n \quad (23)$$

We can show likewise from the properties of the Levi-Civita symbol in n dimensions that

$$\hat{r}_k \cdot \hat{r}_n = 0, \quad k = 1, \dots, n-1 \quad (24a)$$

$$\hat{s}_k \cdot \hat{s}_n = 0, \quad k = 1, \dots, n-1 \quad (24b)$$

$$\hat{r}_n \cdot \hat{r}_n = \hat{s}_n \cdot \hat{s}_n = 1 \quad (25)$$

It follows that the two $n \times n$ matrices R and S defined according to their columns by

$$R = [\hat{r}_1, \hat{r}_2, \dots, \hat{r}_n], \quad S = [\hat{s}_1, \hat{s}_2, \dots, \hat{s}_n] \quad (26)$$

are each proper orthogonal and

$$S = AR \quad (27)$$

Thus,

$$A = SR^T \quad (28)$$

is proper orthogonal and is the desired $n \times n$ attitude matrix.

Comparison with the Triad Method

This method bears a clear resemblance to the triad method^{5,6} first published by Black⁵ in 1964 to solve this particular attitude estimation problem in three dimensions. The resemblance is made all the stronger if we note that Eq. (17) can be written as

$$(\hat{r}_n)_t = \det |\hat{r}_1 \hat{r}_2 \dots \hat{r}_{n-1} \mathbf{1}_t| \quad (29)$$

where $\mathbf{1}_t$ denotes a column vector every element of which vanishes except the t th, which is unity. Equation (29) is the generalization in n dimensions of the vector product, which is prominent in the triad method. The triad algorithm, in fact, had already been in use in the previous decade,¹⁰ and Eq. (28) as the relation between a rotation matrix and the two orthonormal bases it connects can be found for three dimensions in dyadic form in the works of Gibbs.¹¹ One can only imagine that if Eq. (11) were set before Gibbs he would have computed the attitude matrix via the triad method.

The triad method has been variously called the algebraic method¹² or the Sun-Mag method, after the once two most commonly used sensors to which this method was applied. In the triad method, the two sets of orthonormal matrices are computed according to

$$\hat{r}'_1 = V_1/|V_1| \quad (30a)$$

$$\hat{r}'_2 = V_1 \times V_2/|V_1 \times V_2| \quad (30b)$$

$$\hat{r}'_3 = \hat{r}'_1 \times \hat{r}'_2 \quad (30c)$$

and correspondingly for \hat{s}'_k , $k = 1, 2, 3$. The primes distinguish the quantities in the triad method from those just defined.

Clearly, \hat{r}'_1 is identical to \hat{r}_1 . From the well-known Grassman identity

$$a \times (b \times c) = (a \cdot c)b - (a \cdot b)c \quad (31)$$

we see that the definition of \hat{r}'_3 is equivalent to

$$r'_3 = -V_2 + (\hat{r}_1 \cdot V_2)\hat{r}'_1 \quad (32a)$$

$$\hat{r}'_3 = r'_3/|r'_3| \quad (32b)$$

Thus,

$$\hat{r}'_3 = -\hat{r}_2 \quad (33)$$

Finally,

$$(\hat{r}'_2)_k = \sum_{i,j} \epsilon_{ijk} (\hat{r}_1)_i (\hat{r}_2)_j \quad (34)$$

so that

$$\hat{r}'_2 = \hat{r}_3 \quad (35)$$

Thus, in the triad method

$$R' = [\hat{r}'_1, \hat{r}'_2, \hat{r}'_3] = [\hat{r}_1, -\hat{r}_3, \hat{r}_2] \quad (36a)$$

$$S' = [\hat{s}'_1, \hat{s}'_2, \hat{s}'_3] = [\hat{s}_1, -\hat{s}_3, \hat{s}_2] \quad (36b)$$

so that

$$R' = R\Theta, \quad S' = S\Theta \quad (37)$$

where

$$\Theta = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix} \quad (38)$$

is evidently proper orthogonal. The triad solution for the attitude matrix is thus,

$$A = S'R'^T = SR^T \quad (39)$$

and is, therefore, a special case of the general n -dimensional method.

Conclusions

The problem of estimating an $n \times n$ proper orthogonal matrix from N linearly independent vector measurements has been considered. It has been shown that a unique solution for the proper orthogonal matrix exists provided that $N = n - 1$. An algorithm for constructing the attitude matrix based on the Gram-Schmidt orthogonalization method was presented. The special case of this algorithm for $n = 3$ was shown to be equivalent to the well-known triad algorithm.

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