

# METHOD OF THE LEAST SQUARED DISTANCES

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## 1. Introduction

One of the most important tasks in 3-dimensional (3D) object location is the solving of a system of linear equations:

$$\begin{cases} (c_{i9} \times \bar{X}_{ai} - c_{i1}) \times X_a + (c_{i10} \times \bar{X}_{ai} - c_{i2}) \times Y_a + (c_{i11} \times \bar{X}_{ai} - c_{i3}) \times Z_a = c_{i4} - \bar{X}_i \\ (c_{i9} \times \bar{Y}_{ai} - c_{i5}) \times X_a + (c_{i10} \times \bar{Y}_{ai} - c_{i6}) \times Y_a + (c_{i11} \times \bar{Y}_{ai} - c_{i7}) \times Z_a = c_{i8} - \bar{Y}_i \\ i = \overline{1, N} \end{cases} \quad (1)$$

Where:

- $A[X_a, Y_a, Z_a]$  – 3D coordinates of the point  $A$  is in the base coordinate system  $OXYZ$
- $\bar{A}_i[\bar{X}_{ai}, \bar{Y}_{ai}]$  – Object's 2D projection coordinates is in the coordinate system  $\bar{O}_i\bar{X}_i\bar{Y}_i\bar{Z}_i$  connected to the sensor # $i$
- $c_{ij}$  – Calibration coefficient
- $N$  – Total number of sensors

The goal is to find 3D coordinates of the point  $A[X_a, Y_a, Z_a]$  from the measured object's 2D projection coordinates on  $N$  sensors.

Equation (1) can be covered by the following general system:

$$\begin{cases} d_{i0} + \sum_{j=1}^M (d_{ij} \times X_j) = 0 \\ i = \overline{1, N} \end{cases} \quad (2)$$

Where:

- $A[X_1, X_2, \dots, X_M]$  – 3D coordinates of the point  $A$  is in the base coordinate system
- $d_{ij}$  – Coefficient
- $N$  – Total number of sensors
- $M$  – Number of unknowns

## 2. Solution

The goal of this method is to find the sum of the least squared distances from an  $M$ -dimensional point to each  $M$ -dimensional plane, which is the solution of the system (2). Each equation of system (2) represents a plane in  $M$ -dimensional space. If no distortions are present, then  $M$ -dimensional planes intersect at one  $M$ -dimensional point. Otherwise the number of intersection points is more than one. The distance from point  $A[X_1, X_2, \dots, X_M]$  to each  $M$ -dimensional plane, determined by measurement number  $i$ , is the following:

$$\left\{ \begin{array}{l} \delta = \frac{d_{i0} + \sum_{j=1}^M (d_{ij} \times X_i)}{\sqrt{\sum_{j=1}^M d_{ij}^2}} \\ i = \overline{1, N} \end{array} \right. \quad (3)$$

The sum of all squared distances is:

$$\sum_{i=1}^N \delta^2 = \sum_{i=1}^N \left( \frac{d_{i0} + \sum_{j=1}^M (d_{ij} \times X_i)}{\sqrt{\sum_{j=1}^M d_{ij}^2}} \right)^2 = \sum_{i=1}^N \left( c_{i0} + \sum_{j=1}^M (c_{ij} \times X_i) \right)^2 = \text{Min} \quad (4)$$

The partial derivatives of (4) generate the following system of linear equations:

$$\frac{\partial \left( \sum_{i=1}^N \delta^2 \right)}{\partial (X_i)} = 2 \times \sum_{i=1}^N \left( \frac{\left( d_{i0} + \sum_{j=1}^M (d_{ij} \times X_i) \right)}{\sqrt{\sum_{j=1}^M d_{ij}^2}} \right)$$

The minimum of sum (4) can only be when all partial derivatives are equal to zero, leading to the following system of  $M$  equations with  $M$  unknowns:

$$\left\{ \begin{array}{l} 2 \times \left( \sum_{i=1}^N \left( c_{i0} + \sum_{j=1}^M (c_{ij} \times X_i) \right) \right) \times \sum_{j=1}^M c_{ij} = 0 \quad \text{or} \quad \left( \sum_{i=1}^N \left( c_{i0} + \sum_{j=1}^M (c_{ij} \times X_i) \right) \right) = 0 \\ i = \overline{1, N} \end{array} \right. \quad (5)$$

Where:

$$c_{i0} = \frac{d_{i0}}{\sqrt{\sum_{j=1}^M d_{ij}^2}} \quad c_{ij} = \frac{\sum_{i=1}^N d_{i0}}{\sqrt{\sum_{j=1}^M d_{ij}^2}}$$

The system of equation (5) can be solved by any common method.

### 3. Summary

The solution derived by the method of “*the least squared sum of distances*” provides improved results for large coefficient spreads, which, unlike common methods, is less dependant and limited by computer resolution.