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MASTER OF SCIENCE THESIS

# STAR-TRACKER PROGRAM FOR CUBESAT SATELLITES

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**Abstract**

Last years directed space industry towards small satellites. Many countries which did not have possibility to enter this branch of industry before now create their own solutions. The goal of this work is to create fully functioning star-tracker program eligible to be used in future satellites as Polish solution of determination of satellite attitude towards Earth. Work contains also description of individual parts and variants of solutions connected with star-tracker.

**Streszczenie**

Ostatnie lata ukierunkowały przemysł kosmiczny na małe satelity. Wiele krajów, które wcześniej nie miały możliwości wejścia w tę gałąź przemysłu, teraz tworzą własne rozwiązania. Celem niniejszej pracy dyplomowej jest stworzenie w pełni działającego programu star-tracker nadającego się do wykorzystania w przyszłych satelitach jako polskie rozwiązanie problemu określania orientacji satelity względem ziemi. Praca zawiera też opis poszczególnych części i wariantów rozwiązania problemów związanych ze star-trackerem.

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## Nomenclature

$\mathbf{b}$	Known directional unit vector in the BODY frame
$\mathbf{I}$	Identity matrix
$\mathbf{M}$	Least squares estimate of rotation matrix
$\mathbf{n}$	Unit vector
$\mathbf{Q}$	Quaternion matrix
$\mathbf{q}$	Unit quaternion
$\mathbf{q}_{vec}$	Vector part of unit quaternion
$\mathbf{r}$	Known directional unit vector in the NED frame
$\mathbf{R}(\cdot)$	Rotation matrix using Euler angles
$\mathbf{R}_n^b$	Rotation matrix representing a rotation from n to b
$\mathbf{S}(\cdot)$	Skew symmetric matrix
$\phi$	Euler angle, roll
$\psi$	Euler angle, yaw
$\theta$	Euler angle, pitch
$q_0$	Scalar part of unit quaternion
$v$	General Euler angle

[1] [2] [3] [4] [5] [6] [7] [8]

# 1 Introduction

## 1.1 Motivation

The goal of this work is to make fully operational star-tracker program, that could be used on Cubesat satellites. Such program could be used on space missions and could start Polish state-of-the-art technology in growing space technology sector.

## 1.2 Outline of thesis

This thesis consists of several chapters. Here they are shortly summarized:

**Chapter 1** serves as introduction to this thesis and describes the motivation and goal of this work. It also describes the background of the topic.

**Chapter 2** describes all the important foundations for the fully understanding given work.

**Chapter 3** is the main part of this thesis. It describes how the star-tracker program works and goes through detailed comparison of different approaches.

**Chapter 4** describes the created prototype of star-tracker in Python language.

**Chapter 5** talks about the implementation of star-tracker on the existing prototype of on-board computer.

**Chapter 6** describes how the finished program is performing.

**Chapter 7** contains conclusions about this work and created star-tracker program.

### 1.3 Cubesat

Cubesat was designed on CalPoly in 1999[9]. Dimensions of satellite are measured in units. Each unit (often described simply as u) can be 10x10x10cm and can weight up to 1.33 kg. Satellites can be 1u, 2u, 3u, 6u or even 12u.

Such small satellites are susceptible to noise from densly packed electronics.

Zdjecie Cubesata

CubeSat missions, goals, what can they be and are used for? Why is it innovative and important?

### 1.4 Means of attitude estimation

There exist many different types of attitude estimation: sun sensors, star-trackers, magnetometers, etc. However star-tracker gives the best possible accuracy for nowadays and is not susceptible to electrical nor magnetic noise.

#### 1.4.1 Megnetometers

#### 1.4.2 Sun sensors

#### 1.4.3 Earth sensors

#### 1.4.4 GPS

#### 1.4.5 Star trackers

[10] [12]



Sensor	Accuracy	Characteristics and Applicability
Magnetometers	1.0°(5000km alt) 5.0°(200 km alt)	Attitude measured relative to Earth's local magnetic field. Magnetic field uncertainties and variability dominate accuracy. Usable only below $\approx 6,000$ km.
Earth sensors	0.05°(GEO) 0.1°(LEO)	Horizon uncertainties dominate accuracy. Highly accurate units use scanning.
Sun sensors	0.01°	Typical field of view $\pm 30^\circ$
Star sensors	2 arc-sec	Typical field of view $\pm 6^\circ$
Gyroscopes	0.001 deg/hr	Normal use involves periodically resetting reference.
Directional antennas	0.01°to 0.5°	Typically 1 of the antenna beamwidth

Table 1: Sensor Accuracy Ranges. Adapted from [11]

## 1.5 On-board computer

This section will describe the on-board computer which was done as part of other thesis.

## 2 Preliminaries

### 2.1 Coordinate frames

#### 2.1.1 ECI frame

The Earth Centered Inertial frame has its x-axis pointing towards the vernal equinox, and its z-axis pointing along the rotation axis of the Earth at some initial time. The y-axis completes a right handed orthogonal coordinate system. The frame's origin is at the center of the Earth. [10]

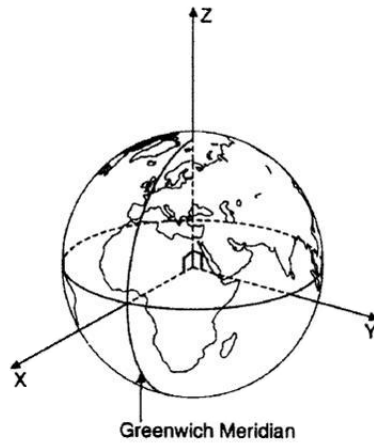


Figure 1: ECI frame, Image [10]

#### 2.1.2 ECEF frame

This frame also has its origin at the center of the Earth, but the Earth Centered Earth Fixed frame has its x-axis pointing towards the point where the intersection between the longitude and latitude have zero value. It can also be described as the intersection between the Greenwich meridian and the Equator. The frame's z-axis is pointing along the Earth's rotation axis.

The y-axis completes the right handed orthogonal system. The ECEF frame is not an inertial frame, it rotates relative to the ECI frame along the Earth rotation.

### 2.1.3 NED frame

The North East Down frame has its z-axis pointing downwards, perpendicular to the tangent plane of the Earth's reference ellipsoid. The ellipsoid is mathematically defined and fitted for approximation of the Earth. The x-axis points towards true north and the y-axis points East. The NED frame is an inertial frame.

### 2.1.4 BODY frame

This frame is attached to the satellite, and is moving and rotating with it. The origin coincides with the origin of the NED frame. The axes coincide with the principle axes of inertia; the x-axis is pointing forwards, the y-axis is pointing to the right side and the z-axis is pointing downwards through the camera side of the satellite.

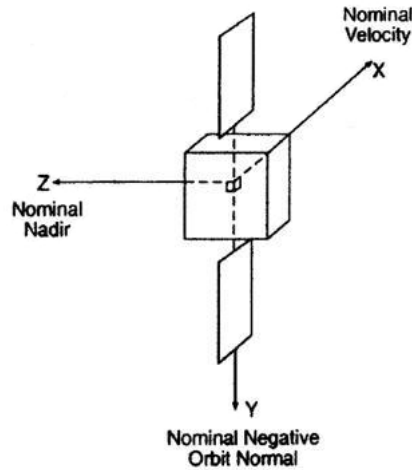


Figure 2: BODY frame, Image [10]

## 2.2 Space environment

## 2.3 Attitude representations

Several representations for describing attitude are available, the most common being Euler angles. More complicated attitude representations are quaternions. Quaternions are used for all the estimation methods presented in this thesis. They are singular-free, and are therefore well suited for attitude determination.

### 2.3.1 Euler angles

[13]

Robot Learning Darmstadt Problems with Euler Angles: Not Unique:  
 Many angles result in the same rotation Hard to quantify differences between  
 two Euler Angles Unit-Quaternion Solves the problems of singularities with  
 the Euler Angles Easier to compute differences of orientations Important if  
 we want to control the orientation of the end-effector See Siciliano or Spong  
 Textbook!

Polar moment

$$\mathbf{R}_x(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos(\phi) & -\sin(\phi) \\ 0 & \sin(\phi) & \cos(\phi) \end{bmatrix} \quad (1)$$

$$\mathbf{R}_y(\theta) = \begin{bmatrix} \cos(\theta) & 0 & \sin(\theta) \\ 0 & 1 & 0 \\ -\sin(\theta) & 0 & \cos(\theta) \end{bmatrix} \quad (2)$$

$$\mathbf{R}_z(\psi) = \begin{bmatrix} \cos(\psi) & -\sin(\psi) & 0 \\ \sin(\psi) & \cos(\psi) & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad (3)$$

### 2.3.2 Quaternions

[14]

[15]

[16]

[17]

[18]

[19]

[20]

$$\mathbf{q} := \begin{bmatrix} q_0 \\ q_1 \\ q_2 \\ q_3 \end{bmatrix} \quad (4)$$

$$q_0 = \cos(v/2) \quad (5)$$

$$\mathbf{n} = \frac{\mathbf{n}}{\|\mathbf{n}\|} \quad (6)$$

$$\mathbf{q}_{vec} := \begin{bmatrix} q_1 \\ q_2 \\ q_3 \end{bmatrix} = [\mathbf{n} \sin(v/2)] \quad (7)$$

$$\mathbf{q} := q_0 + \mathbf{q}_{vec} = q_0 + q_1i + q_2j + q_3k \quad (8)$$

$$\mathbf{Q} = \begin{bmatrix} q_0 & -q_1 & -q_2 & -q_3 \\ q_1 & q_0 & -q_3 & q_2 \\ q_2 & q_3 & q_0 & -q_1 \\ q_3 & -q_2 & q_1 & q_0 \end{bmatrix} \quad (9)$$

$$\mathbf{q}^* := q_0 - \mathbf{q}_{vec} = q_0 + q_1i + q_2j + q_3k \quad (10)$$

$$\mathbf{q}^T \mathbf{q} = 1 \tag{11}$$

## 2.4 Quaternion properties

### 2.4.1 Advantages of quaternions

### 2.4.2 Multiplication of quaternions

### 2.4.3 Quaternions and rotations

## 2.5 Cholesky factorization

## 2.6 Lyapunov analysis

### 3 Star-tracker program

[21]

Generally star-tracker is divided into three main parts[22]:

- recogiting stars on the image and converting the data into list of star vectors by calculating star centroids;
- identyfing which star vector represents which real star in catalogue. This is done by comparing star vectors from the image with data in star catalogue, which is generated before space mission;
- estimating the attitude by calculating the displacement between two frames.

#### 3.1 Centroid - start recognition

[23]

[24]

Due to limitations of camera there exists necessity of calculating star centroids. Each camera converts image into photo divided by pixels. As it is necessary to have high precision of star coordinates, the pixel accuracy is not enough. Subpixel accuracy is needed. Typically it is done by defocusing the lens of the camera and calculating the lumosity of all pixels around the lightest ones. The idea of how to calculate such centroids is adapted from[22].

If FOV is too small, one star will be considered by program as few stars, and if FOV is too large, few stars placed near each other will be considered as one star. Calculating star centroids is tradeoff between counting few stars as one and counting one star as a few. It seems however that it is worse to count one star as few than few stars as one.

$$x_{start} = x - \frac{a_{ROI} - 1}{2} \quad (12)$$

$$y_{start} = y - \frac{a_{ROI} - 1}{2} \quad (13)$$

$$x_{end} = x_{start} + a_{ROI} \quad (14)$$

$$y_{end} = y_{start} + a_{ROI} \quad (15)$$

$$I_{bottom} = \sum_{i=1}^{x_{end}-1} I(i, y_{start}) \quad (16a)$$

$$I_{top} = \sum_{i=2}^{x_{end}} I(i, y_{end}) \quad (16b)$$

$$I_{left} = \sum_{j=1}^{y_{end}-1} I(x_{start}, j) \quad (16c)$$

$$I_{right} = \sum_{j=2}^{y_{end}} I(x_{start}, j) \quad (16d)$$

$$I_{border} = \frac{I_{top} + I_{bottom} + I_{left} + I_{right}}{4(a_{ROI} - 1)} \quad (16e)$$

$$\tilde{I}(x, y) = I(x, y) - I_{border} \quad (17)$$

$$B = \sum_{i=x_{start}+1}^{x_{end}-1} \sum_{j=y_{start}+1}^{y_{end}-1} \tilde{I}(i, j) \quad (18)$$

$$x_{CM} = \sum_{i=x_{start}+1}^{x_{end}-1} \sum_{j=y_{start}+1}^{y_{end}-1} \frac{i \times \tilde{I}(i, j)}{B} \quad (19)$$

$$x_{CM} = \sum_{i=x_{start}+1}^{x_{end}-1} \sum_{j=y_{start}+1}^{y_{end}-1} \frac{j \times \tilde{I}(i, j)}{B} \quad (20)$$

$$u = \frac{\begin{bmatrix} \mu x_{CM} & \mu y_{CM} & f \end{bmatrix}^T}{\| \begin{bmatrix} \mu x_{CM} & \mu y_{CM} & f \end{bmatrix} \|} \quad (21)$$



### 3.2 Star identification

all [25]

Brightness Independent 4-Star Matching Algorithm for Lost-in-Space 3-Axis Attitude Acquisition[26]

SP-Search: A New Algorithm for Star Pattern Recognition [27]

Star Identification using Neural networks [28] [29]

Star pattern recognition using neural networks [30]

#### 3.2.1 Angle Matching

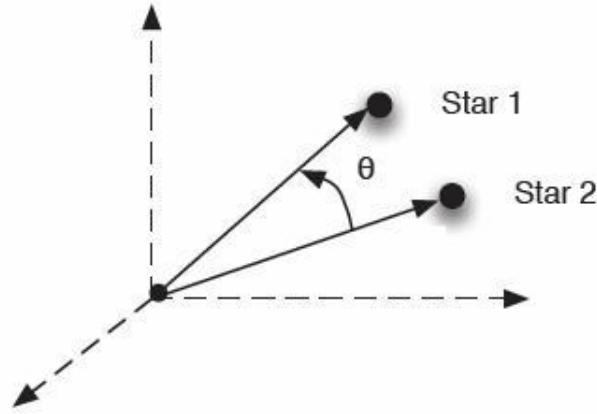


Figure 3: Vector angle method, Image [31]?

[31]

$$\theta = \cos^{-1}(\mathbf{r}_1 \cdot \mathbf{r}_2) \quad (22)$$

$$\mathbf{b}_i = A\mathbf{r}_i \quad (23)$$

$$\tilde{\mathbf{b}}_i = A\mathbf{r}_i + \mathbf{v}_i, \quad \mathbf{v}_i^T A\mathbf{r}_i = 0 \quad (24)$$

$$E \{ \mathbf{v}_i \} = 0 \quad (25a)$$

$$E \{ \mathbf{v}_i \mathbf{v}_i^T \} = \sigma_i^2 [\mathbf{I} - (A\mathbf{r}_i)(A\mathbf{r}_i)^T] \quad (25b)$$

$$\mathbf{b}_i^T \mathbf{b}_j = \mathbf{r}_i^T A^T A \mathbf{r}_j = \mathbf{r}_i^T \mathbf{r}_j \quad (26)$$

$$\tilde{\mathbf{b}}_i = A\mathbf{r}_i + \mathbf{v}_i$$

$$\tilde{\mathbf{b}}_j = A\mathbf{r}_j + \mathbf{v}_j$$

$$z \equiv \tilde{\mathbf{b}}_i^T \tilde{\mathbf{b}}_j = \mathbf{r}_i^T \mathbf{r}_j + \mathbf{r}_i^T A^T \mathbf{v}_j + \mathbf{r}_j^T A^T \mathbf{v}_i + \mathbf{v}_i^T \mathbf{v}_j \quad (28)$$

$$E\{z\} = \mathbf{r}_i^T \mathbf{r}_j \quad (29)$$

$$p \equiv z - E\{z\} = \mathbf{r}_i^T A^T \mathbf{v}_j + \mathbf{r}_j^T A^T \mathbf{v}_i + \mathbf{v}_i^T \mathbf{v}_j \quad (30)$$

$$\begin{aligned} \sigma_p^2 \equiv E\{p\} = \\ \mathbf{r}_1^T A^T R_2 A \mathbf{r}_1 + \mathbf{r}_2^T A^T R_a A \mathbf{r}_2 + \text{Trace}(R_1 R_2) = \\ \text{Trace}(A \mathbf{r}_1 \mathbf{r}_1^T R_2) + \text{Trace}(A \mathbf{r}_2 \mathbf{r}_2^T R_1) + \text{Trace}(R_1 R_2) \end{aligned} \quad (31)$$

### 3.2.2 Spherical Triangle Matching

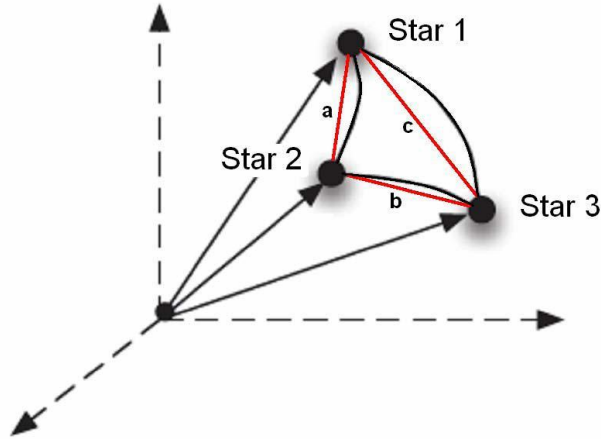


Figure 4: Spherical Triangle Method, Image [32]?

[32]

$$A = 4 \tan^{-1} \sqrt{\tan \frac{s}{2} \tan \frac{s-a}{2} \tan \frac{s-b}{2} \tan \frac{s-c}{2}} \quad (32)$$

$$\begin{aligned}
s &= \frac{1}{2}(a + b + c) \\
a &= \cos^{-1} \left( \frac{b_1 \cdot b_2}{|b_1||b_2|} \right) \\
b &= \cos^{-1} \left( \frac{b_2 \cdot b_3}{|b_2||b_3|} \right) \\
c &= \cos^{-1} \left( \frac{b_3 \cdot b_1}{|b_3||b_1|} \right) \\
I_p &= \sum \theta^2 dA
\end{aligned} \tag{34}$$

### 3.2.3 Planar Triangle

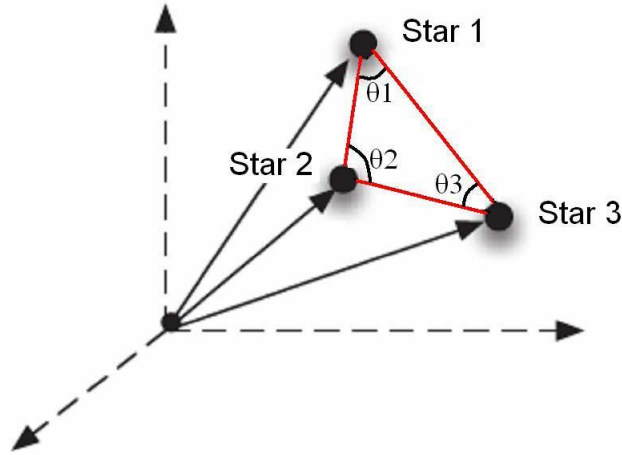


Figure 5: Planar Triangle Method, Image [33]?

[33]

$$s = \frac{1}{2}(a + b + c) \tag{35a}$$

$$a = ||\mathbf{u}_p - \mathbf{u}_q|| \tag{35b}$$

$$b = ||\mathbf{u}_q - \mathbf{u}_r|| \tag{35c}$$

$$c = ||\mathbf{u}_p - \mathbf{u}_r|| \tag{35d}$$

$$A = \sqrt{s(s-a)(s-b)(s-c)} \tag{36}$$

$$J = A \frac{(a^2 + b^2 + c^2)}{36} \quad (37)$$

Derivatives

$$H = \begin{bmatrix} \mathbf{h}_1^T & \mathbf{h}_2^T & \mathbf{h}_3^T \end{bmatrix} \quad (38)$$

$$\mathbf{h}_1^T \equiv \frac{\delta A}{\delta a} \frac{\delta a}{\delta \mathbf{b}_1} + \frac{\delta A}{\delta c} \frac{\delta c}{\delta \mathbf{b}_1} \quad (39a)$$

$$\mathbf{h}_2^T \equiv \frac{\delta A}{\delta a} \frac{\delta a}{\delta \mathbf{b}_2} + \frac{\delta A}{\delta b} \frac{\delta b}{\delta \mathbf{b}_2} \quad (39b)$$

$$\mathbf{h}_3^T \equiv \frac{\delta A}{\delta b} \frac{\delta b}{\delta \mathbf{b}_3} + \frac{\delta A}{\delta c} \frac{\delta c}{\delta \mathbf{b}_3} \quad (39c)$$

$$\frac{\delta A}{\delta a} = \frac{u_1 - u_2 + u_3 + u_4}{4A} \quad (40a)$$

$$\frac{\delta A}{\delta b} = \frac{u_1 + u_2 - u_3 + u_4}{4A} \quad (40b)$$

$$\frac{\delta A}{\delta c} = \frac{u_1 + u_2 + u_3 - u_4}{4A} \quad (40c)$$

$$u_1 = (s - a)(s - b)(s - c) \quad (41a)$$

$$u_2 = s(s - b)(s - c) \quad (41b)$$

$$u_3 = s(s - a)(s - c) \quad (41c)$$

$$u_4 = s(s - a)(s - b) \quad (41d)$$

$$\frac{\delta a}{\delta \mathbf{b}_1} = (\mathbf{b}_1 - \mathbf{b}_2)^T / a, \quad \frac{\delta a}{\delta \mathbf{b}_2} = -\frac{\delta a}{\delta \mathbf{b}_1} \quad (42a)$$

$$\frac{\delta b}{\delta \mathbf{b}_2} = (\mathbf{b}_2 - \mathbf{b}_3)^T / b, \quad \frac{\delta b}{\delta \mathbf{b}_3} = -\frac{\delta b}{\delta \mathbf{b}_2} \quad (42b)$$

$$\frac{\delta c}{\delta \mathbf{b}_1} = (\mathbf{b}_1 - \mathbf{b}_3)^T / c, \quad \frac{\delta c}{\delta \mathbf{b}_3} = -\frac{\delta c}{\delta \mathbf{b}_1} \quad (42c)$$

$$\sigma_A^2 = H R H^T \quad (43)$$

$$R \equiv \begin{bmatrix} R_1 & 0_{3 \times 3} & 0_{3 \times 3} \\ 0_{3 \times 3} & R_2 & 0_{3 \times 3} \\ 0_{3 \times 3} & 0_{3 \times 3} & R_3 \end{bmatrix} \quad (44)$$

Polar Moment

$$\bar{H} = \begin{bmatrix} \bar{\mathbf{h}}_1^T & \bar{\mathbf{h}}_2^T & \bar{\mathbf{h}}_3^T \end{bmatrix} \quad (45)$$

$$\bar{\mathbf{h}}_1^T \equiv \frac{\delta J}{\delta a} \frac{\delta a}{\delta \mathbf{b}_1} + \frac{\delta J}{\delta c} \frac{\delta c}{\delta \mathbf{b}_1} + \frac{\delta J}{\delta A} \mathbf{h}_1^T \quad (46a)$$

$$\bar{\mathbf{h}}_2^T \equiv \frac{\delta J}{\delta a} \frac{\delta a}{\delta \mathbf{b}_2} + \frac{\delta J}{\delta b} \frac{\delta b}{\delta \mathbf{b}_2} + \frac{\delta J}{\delta A} \mathbf{h}_2^T \quad (46b)$$

$$\bar{\mathbf{h}}_3^T \equiv \frac{\delta J}{\delta b} \frac{\delta b}{\delta \mathbf{b}_3} + \frac{\delta J}{\delta c} \frac{\delta c}{\delta \mathbf{b}_3} + \frac{\delta J}{\delta A} \mathbf{h}_3^T \quad (46c)$$

$$\frac{\delta J}{\delta a} = Aa/18, \quad \frac{\delta J}{\delta a} = Ab/18, \quad \frac{\delta J}{\delta a} = Ac/18 \quad (47a)$$

$$\frac{\delta J}{\delta A} = (a^2 + b^2 + c^2)/36 \quad (47b)$$

$$\sigma_J^2 = \bar{H} R \bar{H}^T \quad (48)$$

### 3.2.4 Pyramid

[34]

### 3.2.5 Rate Matching

[35] to be removed?

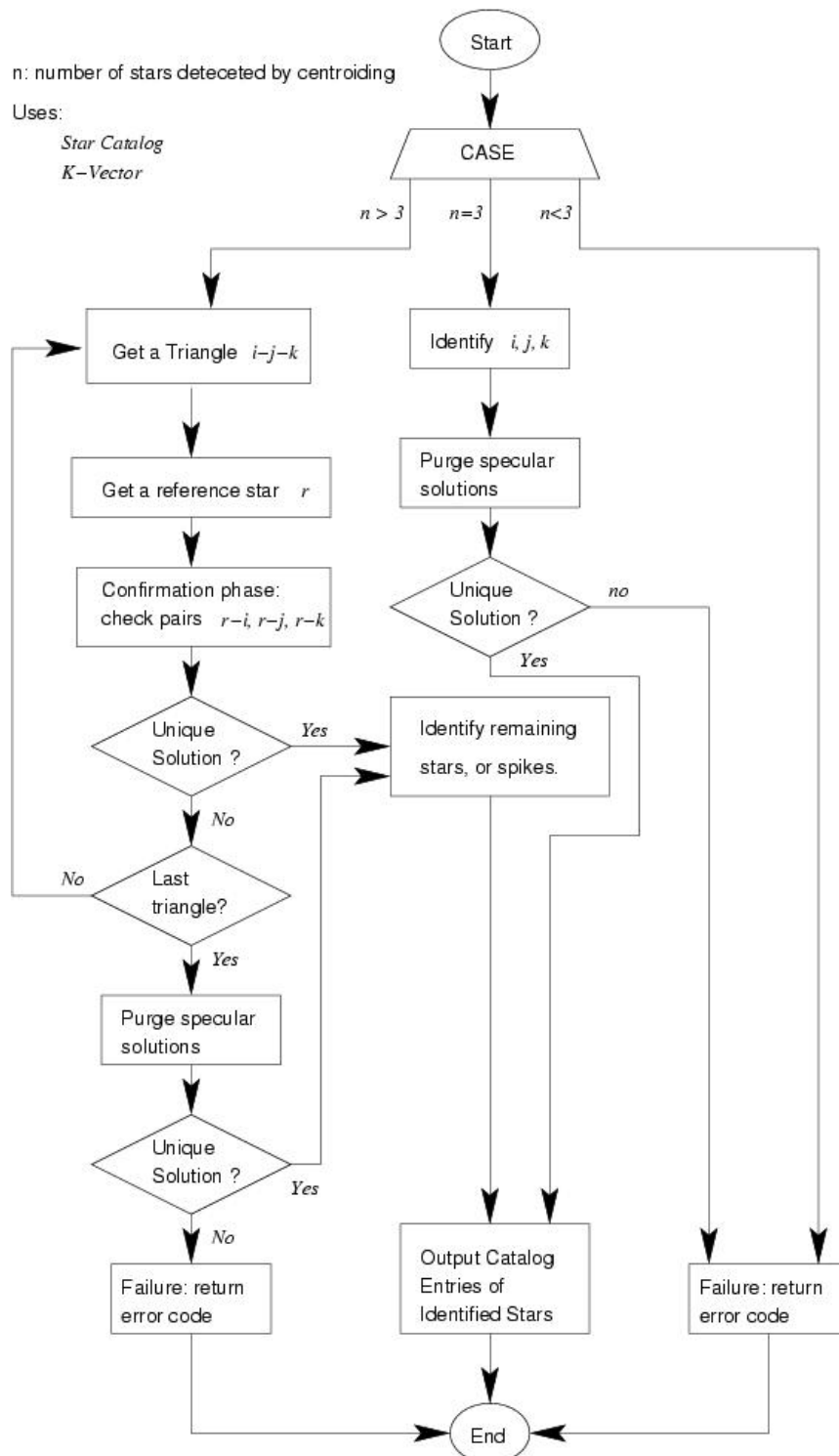


Figure 6: Pyramid Method Flowchart, Image [34]

### 3.2.6 Voting

[36]

### 3.2.7 Grid

[37]

## 3.3 Star-catalogue and searching for matching stars

### 3.3.1 Star Catalogue Generation

$$\mathbf{u} = \begin{bmatrix} \cos \alpha \cos \delta \\ \sin \alpha \cos \delta \\ \sin \delta \end{bmatrix} \quad (49)$$

$$m_i \leq m_{max} \quad (50)$$

$$m_j \leq m_{max} \quad (51)$$

$$\mathbf{u}_a^T \mathbf{u}_b \geq \cos \theta_{FOV} \quad (52)$$

### 3.3.2 Candidate Matching

to be removed?

### 3.3.3 Result Verification

to be removed?

### 3.3.4 k-vector

The k-vector database is built a priori for some given working magnitude threshold and for the star tracker maximum angular aperture. Essentially, the k-vector table is a structural database of all cataloged star pairs that could possibly fit in the camera FOV over the whole sky. The star pairs are ordered with increasing interstar angle. The data stored are the k index, the cosine of the interstar angle, and the master catalog indices I[k] and J[k] of the kth star pair. The k-vector access logic is invoked in real time for a minimal set of star pairs in elementary measured star polygons (three for a triangle, six for a four-star pyramid, etc.); the fact that the vertices between adjacent measured star pairs share a common cataloged star is the key observation leading to logic for efficiently identifying the stars by simply comparing the k-vector accessed catalog indices from the several sets of candidate star pairs (which must contain the common measured pivot star, if it is in the catalog). [38]

[39]

[40]

Trzeba dodać pogrubienia vectorów

$$z(x) = mx + q \quad (53)$$

$$m = \frac{y_{max} - y_{min} + \delta\epsilon}{n - 1} \quad (54)$$

$$q = y_{min} - m - \delta\epsilon \quad (55)$$

$$\epsilon \approx 22.2 \times 10^{-16} \quad (56)$$

$$\delta\epsilon = (n - 1)\epsilon \quad (57)$$

$$k(i) = j \quad \text{where} \quad s(j) \leq z(i) < s(j + 1) \quad (58)$$

or

$$k(i) = j \quad \text{where } j \text{ is the greatest index such } s(j) \leq y(I(i)) \quad \text{is satisfied.} \quad (59)$$

$$j_b = \left\lfloor \frac{y_a - q}{m} \right\rfloor \quad \text{and} \quad j_t = \left\lceil \frac{y_b - q}{m} \right\rceil \quad (60)$$

$$k_{start} = k(j_b) + 1 \quad \text{and} \quad k_{end} = k(j_t) \quad (61)$$



### 3.4 Attitude Determination

[1]

AIM (Attitude estimation using Image Matching)[3]

all [11] [41]

#### 3.4.1 The Predictive Attitude Determination Algorithm ?

[42]

#### 3.4.2 q-method

$$\mathbf{s}_b = \mathbf{R}^{bi} \mathbf{s}_i \quad \mathbf{m}_b = \mathbf{R}^{bi} \mathbf{m}_i \quad (62)$$

$$\begin{aligned} J &= \frac{1}{2} \sum w_k (\mathbf{v}_{kb} - \mathbf{R}^{bi} \mathbf{v}_{ki})^T (\mathbf{v}_{kb} - \mathbf{R}^{bi} \mathbf{v}_{ki}) \\ &= \frac{1}{2} \sum w_k (\mathbf{v}_{kb}^T \mathbf{v}_{kb} + \mathbf{v}_{ki}^T \mathbf{v}_{ki} + 2 \mathbf{v}_{kb}^T \mathbf{R}^{bi} \mathbf{v}_{ki}) \end{aligned} \quad (63)$$

$$J = \sum w_k (1 - \mathbf{v}_{kb}^T \mathbf{R}^{bi} \mathbf{v}_{ki}) \quad (64)$$

$$g(\mathbf{R}) = \sum w_k \mathbf{v}_{kb}^T \mathbf{R}^{bi} \mathbf{v}_{ki} \quad (65)$$

$$\mathbf{R} = (q_4^2 - \mathbf{q}^T \mathbf{q}) \mathbf{1} + 2 \mathbf{q} \mathbf{q}^T - 2 q_4 \mathbf{q}^x \quad (66)$$

$$\bar{\mathbf{q}}^T \bar{\mathbf{q}} = 1 \quad (67)$$

$$g(\bar{\mathbf{q}}) = \bar{\mathbf{q}}^T \mathbf{K} \bar{\mathbf{q}} \quad (68)$$

$$\mathbf{K} = \begin{bmatrix} \mathbf{S} - \sigma \mathbf{I} & \mathbf{Z} \\ \mathbf{Z}^T & \sigma \end{bmatrix} \quad (69)$$

$$\mathbf{B} = \sum_{k=1}^N w_k (\mathbf{v}_{kb} \mathbf{v}_{ki}^T) \quad (70)$$

$$\mathbf{S} = \mathbf{B} + \mathbf{B}^T \quad (71)$$

$$\mathbf{Z} = \begin{bmatrix} B_{23} - B_{32} & B_{32} - B_{13} & B_{12} - B_{21} \end{bmatrix}^T \quad (72)$$

$$\sigma = \text{tr}[\mathbf{B}] \quad (73)$$

$$g'(\bar{\mathbf{q}}) = \bar{\mathbf{q}}^T \mathbf{K} \bar{\mathbf{q}} - \lambda \bar{\mathbf{q}}^T \bar{\mathbf{q}} \quad (74)$$

$$\mathbf{K} \bar{\mathbf{q}} = \lambda \bar{\mathbf{q}} \quad (75)$$

$$g(\bar{\mathbf{q}}) = \bar{\mathbf{q}}^T \mathbf{K} \bar{\mathbf{q}} = \bar{\mathbf{q}}^T \lambda \bar{\mathbf{q}} = \lambda \bar{\mathbf{q}}^T \bar{\mathbf{q}} = \lambda \quad (76)$$

### 3.4.3 Wahba's problem

[43]

$$\sum_j^n ||r_j - Mb_j|| \quad (77)$$

### 3.4.4 QUEST

improvement to quest implementation [44]

kallman filtering [45]

$$J(\mathbf{q}) = \frac{1}{2} \sum_{j=1}^n \frac{1}{\sigma_j^2} (\mathbf{b}_j - \mathbf{R}_b^i(\mathbf{q}) \mathbf{r}_j)^T (\mathbf{b}_j - \mathbf{R}_b^i(\mathbf{q}) \mathbf{r}_j) =$$

$$\frac{1}{2} \sum_{j=1}^n \frac{1}{\sigma_j^2} (\mathbf{b}_j^T \mathbf{b}_j - 2 \mathbf{b}_j^T \mathbf{R}_b^i(\mathbf{q}) \mathbf{r}_j + \mathbf{r}_j^T \mathbf{r}_j)$$
(78)

$$J(\mathbf{q}) = \sum_{j=1}^n \frac{1}{\sigma_j^2} (1 - \mathbf{b}_j^T \mathbf{R}_b^i(\mathbf{q}) \mathbf{r}_j)$$
(79)

### 3.4.5 TRIAD

a must

### 3.4.6 The Fast Optimal Attitude Matrix

to be removed?

### 3.4.7 DCM (Direction Cosine Matrix)

[46] and

[22]

$$\mathbf{B} = \sum_{i=1}^n \mathbf{b}_i \mathbf{r}_i^T$$
(80)

$$\mathbf{B} = \mathbf{U} \mathbf{S} \mathbf{V}^T$$
(81)

$$\mathbf{U}_+ = \mathbf{U} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \det \mathbf{U} \end{bmatrix}$$
(82)

$$\mathbf{V}_+ = \mathbf{V} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \det \mathbf{V} \end{bmatrix} \quad (83)$$

$$\mathbf{A} = \mathbf{U}_+ \mathbf{V}_+^T \quad (84)$$

## 4 Prototype

For now the following parts are finished in Python:

1. Centroiding
2. Planar Triangle Recognition with variations (nearly)
3. Pyramid alg ?
4. k-vector
5. QUEST (not started yet)

Testing

[47]

## 5 Complete program

## **6    Testing of star-tracker**

[48]

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