



# Full State Pose Estimation Using a Satellite Imager

by

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*To God, my Wife and my Mother*

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- \* Prof. H.W. Jordaan, thank you for all your patience and guidance through this journey.
- \* Pinkmatter for your financial support for taking the risk of investing in me and my future. I will be eternally grateful. Thank you for letting me feel like part of the company and taking such good care of me during the visits.
- \* Clarissa, my wife, thank you for all your love and support during the difficult times.
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- \* God, for all your guidance, support, love and faith in me, even though I didn't deserve it.

*“ Above all else remember the friends you made along the way,  
because it is not your journey that defines you, it is the people you help  
and help you along the way. ”*

– Mr. Niel Theron

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# DECLARATION

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# ABSTRACT

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Pose estimation on nanosatellites is still an on going topic of interest. It is important for satellite to know there position and attitude to do accurate target tracking. Traditional solutions to the pose estimation problem is mainly star trackers, which looks at the constalations of stars to determine the attitude and GPS to determine the position of the satellite along with other sensors like magnetometers and coarse sun sensors.

In this thesis, a sensor is developed that utilises the onboard satellite imager, to estimate the position and the attitude of the satellite. The sensor uses a camera model to take pictures of the Earth surface, a feature detector is ran on the image using scale invariant feature transform (SIFT) to identify and establish corospondance of features. A full state kinematic estimator using the extended Kalman Filter (EKF) based on the simultaneous localisation and mapping (SLAM) approach. The filter makes used of feature vectors and feature discriptors detected on the image. This is used to estimate attitude and position of the satellite.

An simulation environment in MATLAB is developed to propagate a satellite and determine the ground truth pose. Several traditional sensors like the star tracker and magnetometer and GPS to be able to compare the Earth Tracker and create the possiblity to fuse the sensors and determine the accuracy. Results show that the filter estimates the system states successfully. It is concluded that ...

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# UITTREKSEL

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# NOMENCLATURE

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## VARIABLES AND FUNCTIONS

### CONSTANTS

$\omega_e$                       Rotation speed of the Earth

$c$                          A constant.

### FUNCTIONS

$f$                          A function.

### VARIABLES

$x$                          A variable.

## ACRONYMS AND ABBREVIATIONS

<b>ADCS</b>	Attitude Determination and Control System
<b>BRF</b>	Body Reference Frame
<b>CRF</b>	Camera Reference Frame
<b>DCM</b>	Direction Cosine Matrix
<b>ECEF</b>	Earth Centred Earth Fixed
<b>ECI</b>	Earth Centred Inertial
<b>EKF</b>	Extended Kalman Filter
<b>GPS</b>	Global Positioning System
<b>LLA</b>	Latitude Longitude and Altitude Reference Frame
<b>LVLH</b>	Local Vertical Local Horizon
<b>SLAM</b>	Simultaneous Localisation and Mapping

## DEFINITIONS

### A

**ATTITUDE** The orientation of a satellite in space.

### P

**POSE** The combination of a satellite's position and attitude.

### S

**STATE ESTIMATION** The ability to determine a state of a system using mathematical models.

**STUDENT** is an entity needing a thesis to transcend the state of being a student.

## VARIABLES AND FUNCTIONS

### CONSTANTS

$\omega_e$                       Rotation speed of the Earth

$c$                          A constant.

### FUNCTIONS

$f$                          A function.

### VARIABLES

$\mathbf{x}$                         A variable.

# INTRODUCTION

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## 1.1 PROBLEM BACKGROUND

- Satellites are getting smaller - Because this leads to satellites having reduced costs and timelines - This is enabled by the miniaturisation of electronics
  - One of the big industries in satellites is remote sensing - Remote Sensing is the application where satellites are used to monitor the Earth - One of the applications is to take images of the Earth
    - This leads to the problem that high accuracy is needed to take images of the targets on the Earth's surface - COTS components which is mainly used on small satellites lack the accuracy needed - Magnetometers is too low of an accuracy - Star Trackers have the right accuracy, but is expensive

## 1.2 PROPOSED SOLUTION

- Proposed solution is to develop an estimation algorithm that can estimate the full state of the satellite - The Full State of a Satellite is its position in Space and its attitude or its orientation in space. - The satellite uses the imager itself to determine position and attitude. - This can lead to reduce costs as the satellite is using an instrument which is already onboard the satellite. - Utilising the components when it is idle - Observing the target directly

## 1.3 DOCUMENT OUTLINE

- Chapter 2: Will investigate previous sensors that is being used to determine Pose - Previous techniques estimating the pose - Some light touching on feature detection as this is crucial to the pose estimation system

- Chapter 3: Will introduce the modelling of the system - Rigid Body Kinematics - Position Kinematics - Attitude Kinematics - Kalman Filters - Extended Kalman Filters
- Chapter 4: Measurement Generation - Feature detection - PinHole Camera Model. - The Plant - The Plant Model - The Measurement Model
- Chapter 5: State estimation - The Extended Kalman Filter - Update Step - Prediction Step - Simulator
- Chapter 6 is results
- Chapter 7 is Conclusion - Future Work



## LITERATURE

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### 2.1 INTRODUCTION

### 2.2 POSITION AND ATTITUDE DETERMINATION

### 2.3 GEOLOCATION

### 2.4 FEATURE DETECTION

### 2.5 ATTITUDE SOLUTIONS

#### 2.5.1 ATTITUDE DETERMINATION

#### 2.5.2 POSITION DETERMINATION

### 2.6 CONCLUSION

# MODELLING

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## 3.1 INTRODUCTION

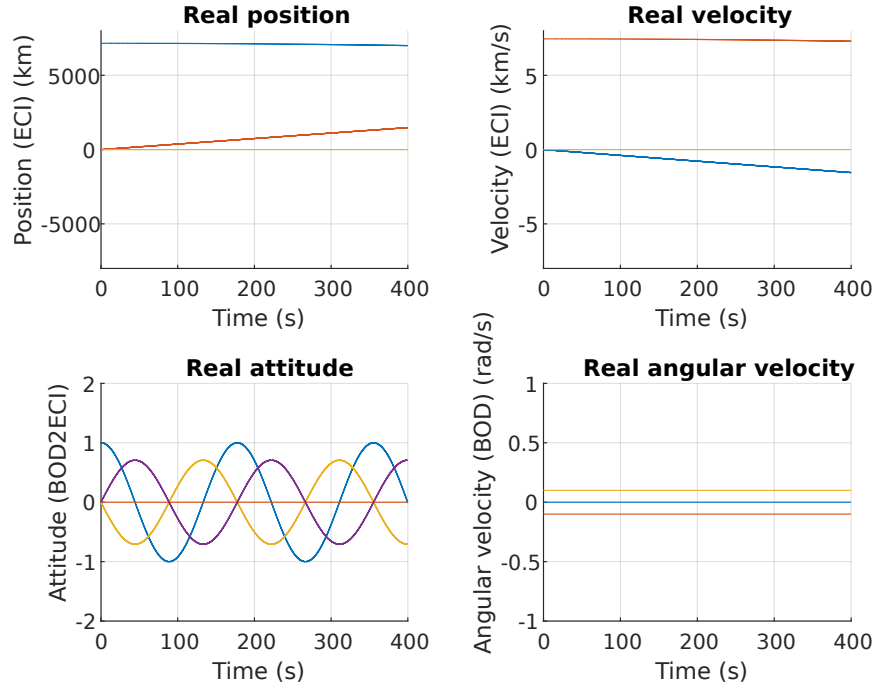
This project focuses on the pose estimation of a satellite using an satellite image. This is essentially a localisation problem and requires a realistic description of the system. The aim of this chapter is to sufficiently define the problem and the proposed solution. Estimation algorithms are discussed and an estimator is chosen to solve the localisation problem. Further, attitude representations of a rigid body are introduced along with the dynamic and kinematic models used to describe a satellite in inertial space. Attention is given to quaternion attitude representations along with their propagation using angular rates.

## 3.2 PROBLEM DEFINITION

The satellite is orbiting in an inertial reference frame (ECI). It has different sensors to estimate the satellite's pose. What is my problem I want to solve. So I have a satellite that is orbiting the earth. This satellite has a camera connected to the body reference frame (BRF) which takes satellite imagery. This image is then passed through a feature detector to determine the features in the system. These features are used to create an internal catalogue which is used for the attitude and position estimation. The feature vector is determined by the camera characteristics which are in the BRF. Therefore the attitude and position dynamics, must be described.

The problem of localising a robot in an unknown environment is often solved using simultaneous localisation and mapping (SLAM). SLAM is a method used by a robot to map an unknown environment and simultaneously locate itself in the map. The sensor, with a pose  $x_t$ , receives measurements  $z_t$ , at a given time step  $t$ . Given the measurements, the aim is to estimate the sensor's location relative to the detected features.

The sensor with the reference frame  $\mathcal{B}$ , shown in Figure 3.1



**Figure 3.1:** Your figure caption.

## 3.3 RIGID BODY MECHANICS

### 3.3.1 KINEMATICS

The pose of a rigid body in a reference frame consists of the position and attitude of the body. The attitude, or orientation of a body-fixed reference frame to a known reference frame. This is usually represented by a rotation matrix, often referred to as a direction cosine Matrix (DCM). A rotation about a single coordinate axis is referred to as a coordinate rotation. A coordinate rotation about the x-,y- and z-axes with angles  $\phi$ ,  $\theta$  and  $\psi$ , of the body can be respectively describes as, [Willem de Jong p.23]

$$R_x(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos(\phi) & \sin(\phi) \\ 0 & -\sin(\phi) & \cos(\phi) \end{bmatrix} \quad (3.1)$$

$$R_y(\theta) = \begin{bmatrix} \cos(\theta) & 0 & -\sin(\theta) \\ 0 & 1 & 0 \\ \sin(\theta) & 0 & \cos(\theta) \end{bmatrix} \quad (3.2)$$

$$R_z(\psi) = \begin{bmatrix} \cos(\phi) & \sin(\phi) & 0 \\ -\sin(\phi) & \cos(\phi) & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad (3.3)$$

Any rotation in 3D space can be described by three coordinate rotations. The DCM describing the attitude of the target in the camera reference frame (CRF),  $\mathbf{A}_C^B$ , can be represented by three Euler angles. Each of the angles corresponds to one coordinate rotation. The order of the Euler 1-2-3 rotation, shown in Figure 3.5, is expressed as

$$\mathbf{A}_C^B = R_x(\phi)R_y(\theta)R_z(\psi) \quad (3.4)$$

$$\begin{bmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \end{bmatrix} \quad (3.5)$$

$$\begin{bmatrix} C\theta C\psi & C\theta S\psi & -S\theta \\ S\phi S\theta C\psi - C\phi S\psi & S\phi S\theta S\psi + C\phi C\psi & S\phi C\theta \\ C\phi S\theta C\psi + S\phi S\psi & C\phi S\theta S\psi - S\phi C\psi & C\phi C\theta \end{bmatrix} \quad (3.6)$$

Where S is the sine function and C is the cosine function. The Euler angles are calculated as follows

$$\phi = \arctan 2 \left( \frac{a_{2,3}}{a_{3,3}} \right) \quad (3.7)$$

$$\theta = \arctan 2 \left( \frac{-a_{1,3}}{\sqrt{a_{1,1}^2} + \sqrt{a_{1,2}^2}} \right) \quad (3.8)$$

$$\psi = \arctan 2 \left( \frac{a_{1,2}}{a_{1,1}} \right) \quad (3.9)$$

mathematical singularities occur when using Euler angles to represent large rotations. When both  $a_{1,1}$  and  $a_{1,2}$  in Equation 3.4 are zero, the expressions for  $\psi$  and  $\theta$  are undefined. This is known as *gimbal lock*, where the changes in the first and third Euler angles are indistinguishable when the second angle nears a critical value. Alternatively, the DCM can be described using quaternions, which do not have these singularities. The quaternion rotation in Figure ?? is expressed by the Euler axis  $\bar{\mathbf{e}} = [e_x, e_y, e_z]^T$  and the angle  $\theta$

$$\mathbf{q} = \begin{bmatrix} q_s \\ q_x \\ q_y \\ q_z \end{bmatrix} = \begin{bmatrix} \cos(\theta/2) \\ e_x \sin(\theta/2) \\ e_y \sin(\theta/2) \\ e_z \sin(\theta/2) \end{bmatrix} \quad (3.10)$$

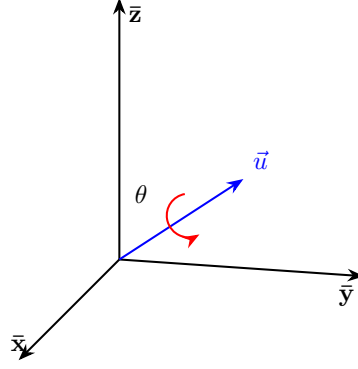


Figure 3.2: Quaternion Rotation

The DCM as a function of Quaternion set is expressed as,

$$\mathbf{A}_C^B = \begin{bmatrix} q_s^2 + q_x^2 - q_y^2 - q_z^2 & 2(q_x q_y - q_s q_z) & 2(q_x q_z + q_s q_y) \\ 2(q_x q_y + q_s q_z) & q_s^2 - q_x^2 + q_y^2 - q_z^2 & 2(q_y q_z - q_s q_x) \\ 2(q_x q_z - q_s q_y) & 2(q_y q_z + q_s q_x) & q_s^2 - q_x^2 - q_y^2 + q_z^2 \end{bmatrix} \quad (3.11)$$

Using the normalisation constraint,  $q_s^2 + q_x^2 + q_y^2 + q_z^2 = 1$ , the DCM Simplifies to,

$$\mathbf{A}_C^B = \begin{bmatrix} 1 - 2(q_y^2 + q_z^2) & 2(q_x q_y - q_s q_z) & 2(q_x q_z + q_s q_y) \\ 2(q_x q_y + q_s q_z) & 1 - 2(q_x^2 + q_z^2) & 2(q_y q_z - q_s q_x) \\ 2(q_x q_z - q_s q_y) & 2(q_y q_z + q_s q_x) & 1 - 2(q_x^2 + q_y^2) \end{bmatrix} \quad (3.12)$$

The body-fixed angular rates of the satellite in CRF,  $\omega_C^B$ , is expressed as a function of quaternions by,

$$\omega_C^B = \begin{bmatrix} \omega_{bx} \\ \omega_{by} \\ \omega_{bz} \end{bmatrix} = 2 \begin{bmatrix} -q_x & q_s & -q_z & q_y \\ -q_3 & q_4 & q_1 & -q_2 \\ -q_4 & -q_3 & q_2 & q_s \end{bmatrix} \begin{bmatrix} \dot{q}_s \\ \dot{q}_x \\ \dot{q}_y \\ \dot{q}_z \end{bmatrix} \quad (3.13)$$

Inversly the quaternion rates as a function of the body rates are,

$$\begin{bmatrix} \dot{q}_s \\ \dot{q}_x \\ \dot{q}_y \\ \dot{q}_z \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 0 & -\omega_{bx} & -\omega_{by} & -\omega_{bz} \\ \omega_{bx} & 0 & \omega_{bz} & -\omega_{by} \\ \omega_{by} & -\omega_{bz} & 0 & \omega_{bx} \\ \omega_{bz} & \omega_{by} & -\omega_{bx} & 0 \end{bmatrix} \begin{bmatrix} q_s \\ q_x \\ q_y \\ q_z \end{bmatrix} \quad (3.14)$$

Quaternions will be used throughout this thesis for attitude representations. Quaternions do not have ambiguity regarding the order of rotations and the rotation is around a well-defined axis. The sin and cosine elements of the rotation matrix are already encoded in the quaternion form of the DCM. Therefore, only one matrix operation is required for attitude transforms, where Euler angles require three.

### 3.3.2 DYNAMICS

## 3.4 REFERENCE FRAME TRANSFORMATIONS

In this masters we are going to encounter a few different reference frames. To accurately create the measurement model we should have an understanding of all the different reference models and how to transform from one to another

### 3.4.0.1 LATITUDE, LONGITUDE AND ALTITUDE

The latitude, longitude of a feature or the position of the satellite is denoted with the  $\mathcal{L}$ . The latitude of a feature is the position of how high or low it is above the equator, having a range of  $-90^\circ$  to  $90^\circ$ . The longitude is based on the Greenwich meridian, a longitude line that passes through the north- and south pole, it has a range of  $-180^\circ$  to  $180^\circ$ . The altitude is measured from the "WGS84" elliptical globe.

[Insert Figure](#)

### 3.4.1 EARTH CENTERED EARTH FIXED

The Earth Centered Earth Fixed reference frame is represented by the  $\mathcal{F}$  and is very similar to the  $\mathcal{L}$  reference frame with the z-axis aligned with the north pole and the x-axis points at the crossing of the Prime Meridian and the Equator, where the y-axis completes the right hand rule. The x, y and z-axis is defined in kilometers. To convert from  $\mathcal{L}$  to  $\mathcal{F}$  is to use a "WGS84" transform. Where WGS84 stands for World Geodetic System 1984, which is the standard coordinate system used for Global Positioning System (GPS). The WGS84 transformation uses a reference ellipsoid that uses a semi-major axis of 6,378 km and a flattening of  $1/298.2$

$$\mathbf{A}_{\mathcal{L}}^{\mathcal{F}} = f(\text{WGS84}) \quad (3.15)$$

[Insert Figure Here](#)

### 3.4.2 EARTH CENTERED INERTIAL

The Earth Centered Inertial reference frame (ECI) referenced by  $\mathcal{I}$  shares a reference frame axis with the ECEF, but is rotated about the z-axis. This rotation is governed by the rotation speed of the earth  $\omega_e$  which is  $7.2921 \times 10^{-5}$  rad/s and time  $t$ .

$$\mathbf{A}_{\mathcal{F}}^{\mathcal{I}} = R(\omega_e t) = \begin{bmatrix} \cos(\omega_e t) & -\sin(\omega_e t) & 0 \\ \sin(\omega_e t) & \cos(\omega_e t) & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad (3.16)$$

Insert Figure Here

### 3.4.3 ORBITAL REFERENCE FRAME

The orbital reference frame used is the Local Vertical Local Horizon (LVLH) denoted by  $\mathcal{O}$ . The LVLH frame is a rotating, orbit-attached coordinate system commonly used in spacecraft dynamics. It moves with the satellite and is defined relative to its orbit around Earth. The x-axis is the "Local Horizon" also called "along track" pointing forward it is tangent to the orbit and points in the direction of motion. The z-axis is the local vertical and is also called the Nadir direction, it points to the barycenter of the system, in this case the center of the Earth. The y-axis is called the cross track it completes the right handed system. It points out of the orbital plane, typically the angular momentum vector direction (normal to the orbit plane).

if  $\mathbf{r}$  is the position vector of the satellite and  $\mathbf{v}$  is the velocity vector of the satellite. The equation for the reference frame is:

$$\bar{z}_{\mathcal{O}} = -\frac{\mathbf{r}}{\|\mathbf{r}\|} \quad (3.17)$$

$$\bar{y}_{\mathcal{O}} = \frac{\mathbf{r} \times \mathbf{v}}{\|\mathbf{r} \times \mathbf{v}\|} \quad (3.18)$$

$$\bar{x}_{\mathcal{O}} = \bar{y}_{\mathcal{O}} \times \bar{z}_{\mathcal{O}} \quad (3.19)$$

For this reference frame there should also be a reference frame translation introduced. Which is done by substracing  $\mathbf{r}$  from the vector

$$\mathbf{f}_{\mathcal{O}} = \mathbf{A}_{\mathcal{I}}^{\mathcal{O}} \times (\mathbf{f}_{\mathcal{I}} - \mathbf{r}_{\mathcal{I}}) \quad (3.20)$$

Insert Figure Here. This is unfinished explain a bit more. Actually want to change it to the 4x4 transformation matrix

### 3.4.4 CAMERA REFERENCE FRAME

The camera reference frame denoted by  $\mathcal{C}$  and the body reference frame  $\mathcal{B}$  in this thesis is the same reference frame. This reference frame is transformed by using your standard quaternion rotation matrix.

$$\mathbf{f}_{\mathcal{C}} = \mathbf{A}_{\mathcal{O}}^{\mathcal{C}} \times \mathbf{f}_{\mathcal{O}} \quad (3.21)$$

## 3.5 STATE ESTIMATION

## 3.6 CONCLUSION



## IMAGE PROCESSING

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### 4.1 INTRODUCTION

### 4.2 PINHOLE CAMERA MODEL

### 4.3 MEASUREMENT EXTRACTION

## STATE ESTIMATION

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### 5.1 INTRODUCTION

### 5.2 EXTENDED KALMAN FILTER

### 5.3 SYSTEM MODELLING

#### 5.3.1 MOTION MODEL

#### 5.3.2 MEASUREMENT MODEL

### 5.4 SIMULATION

### 5.5 PRACTICAL CONSIDERATION

#### 5.5.1 NUMBER OF FEATURES

#### 5.5.2 OUTLIERS

### 5.6 CONCLUSION

# EXPERIMENTS

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6.1 INTRODUCTION

6.2 CONCLUSION

## CONCLUSION AND FUTURE WORK

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### 7.1 CONCLUSION

### 7.2 FUTURE WORK

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## REFERENCES

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## APPENDIX A

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APPENDIX TITLE GOES HERE

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