





# **Research Day**

Verification of reliability of multi-UAV for their use in any scenario

Project sub-group:

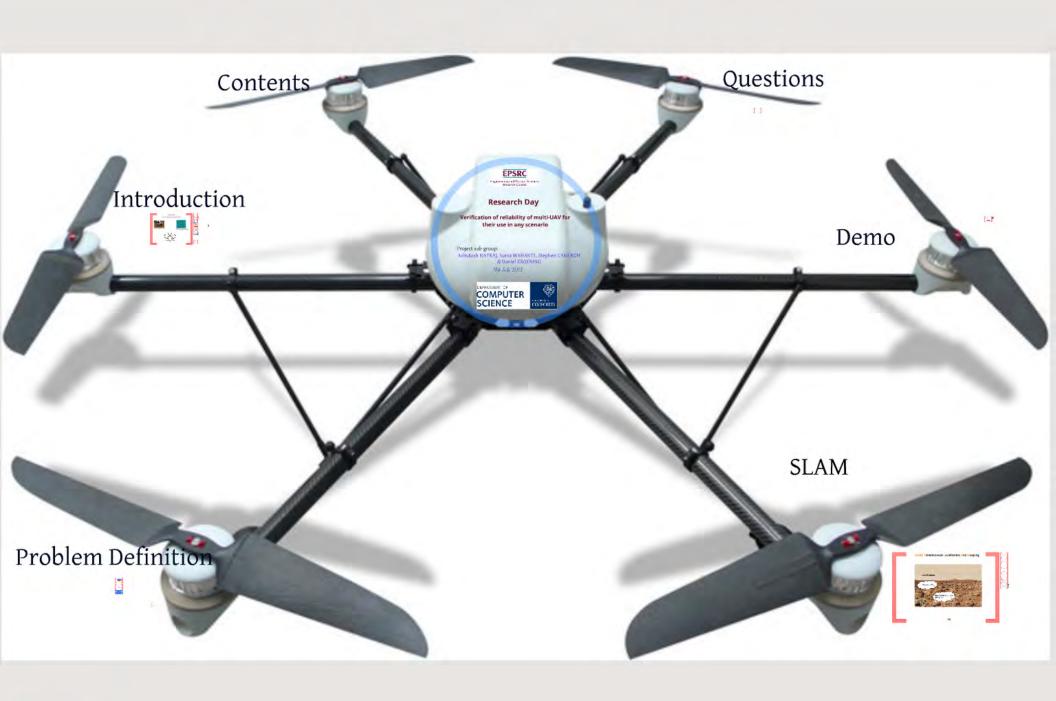
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11th July 2013

COMPUTER SCIENCE





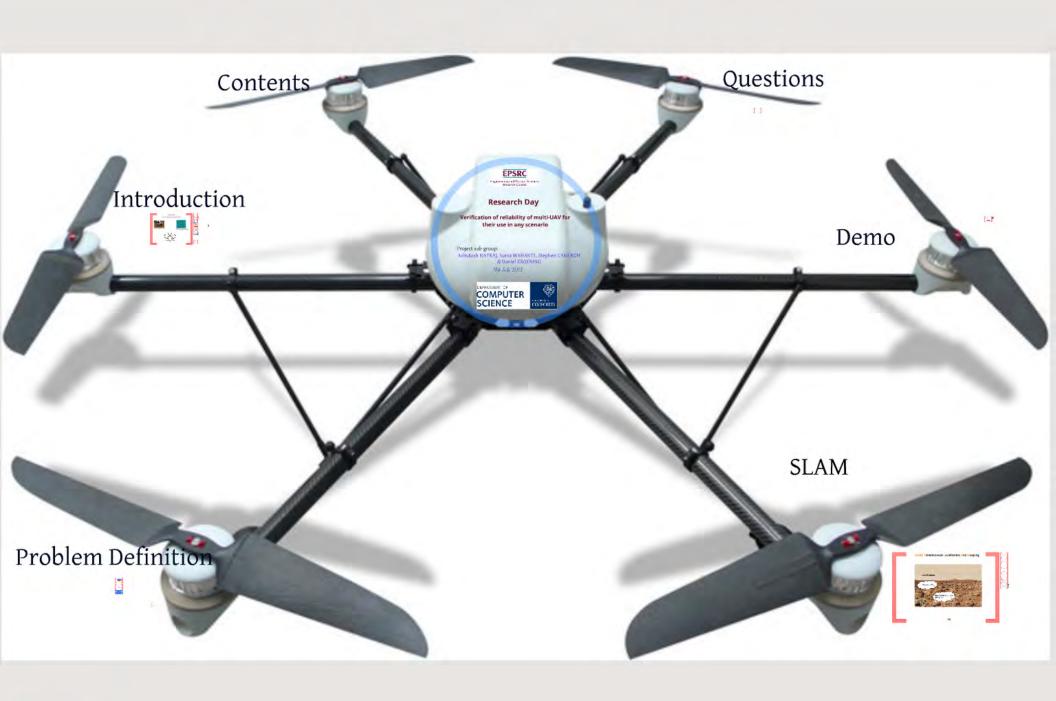




## Contents

Introduction	2
Problem Definition	10
SLAM	13
Experiment Demo	21
Questions	22
Live Demo	





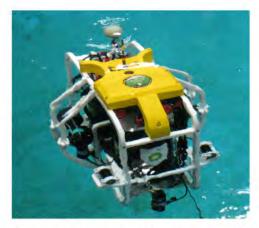


### Introduction

### Types of Autonomous Unmanned Vehicles



Unmanned Ground Vehicles (UGVs)



Autonomous Underwater Vehicles (AUVs)



Unmanned Aerial Vehicles (UAVs)

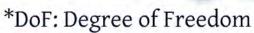


### Introduction











6 DoF, one of the most complicated but also most versatile





# **Types of UAVs**





**Fixed Wing UAV** 







Rotor based UAVs: Can operate in limited space & hover.

### Reliability in Robotic Applications is needed

Aerial Photography Cui et.al (2008) low cost, reliable, fast and hassle free.



Fukushima disaster aerial monitoring

UAV for football match/ movie recording



Highway traffic monitoring Ro et.al (2007) help reduce bottle neck traffic congestion.

Monitoring traffic offenders Puri et.al (2008) monitoring traffic offenders & collecting traffic data.



Aerial traffic monitoring & Bridge inspection
Source: The University of Minnesota, Aerospace Engineering, website.

Wild forest fire monitoring.



Coastal & Marine life research Myers et.al (2005)

Wild forest fire Merino et.al (2008)

Planning military coup operations. Used against the militants.



### **Important Challenges For UAVs**

Take Off and Landing of UAVs

Control & Navigation of UAVs

Essentially Require Knowledge About:

Attitude

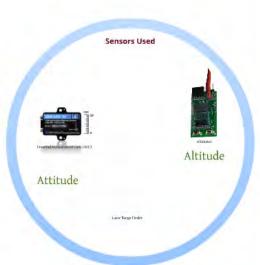
Altitude

Pitch (rotation along Y axis)

Roll (rotation along X axis)



6







#### **Sensors Used**



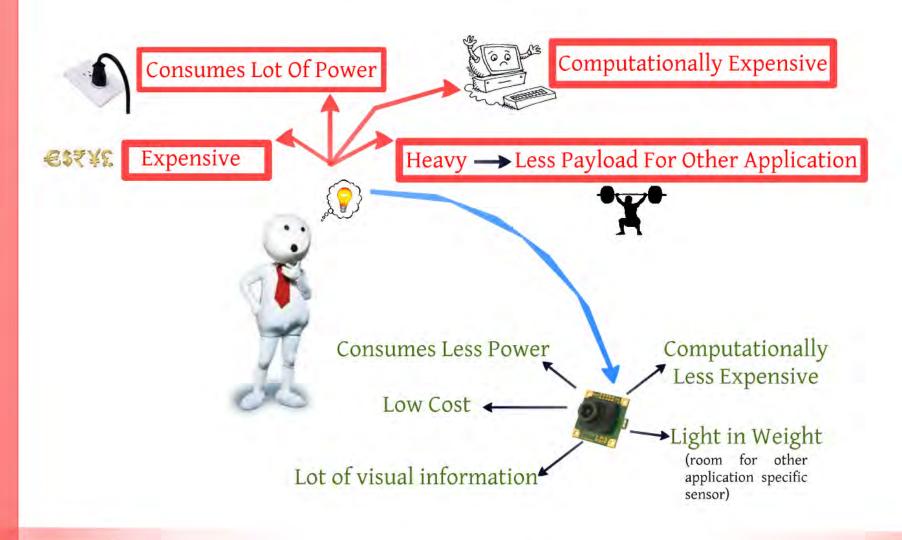


Altimeter

# Altitude



### **Drawbacks Of Using Multiple Sensors**





### **Vision Based Tools For Robotic Applications**

#### **Tools**

### **Applications**

Visual Odometry



Maimone et.al (2007)

Visual Servoing



Hutchison et.al (1996)

Visual SLAM



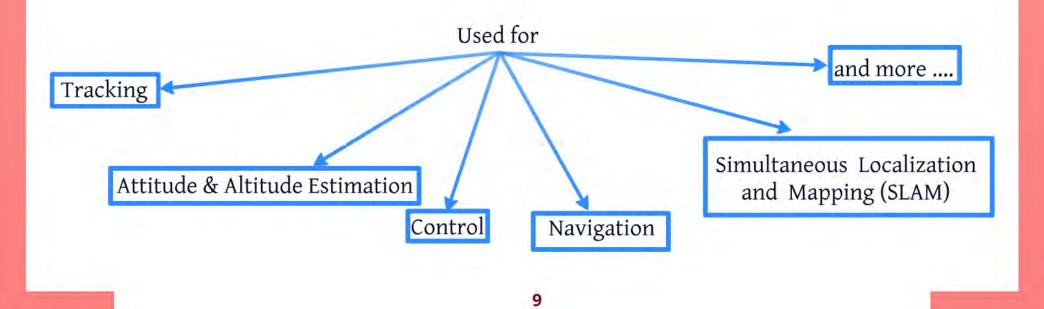
Kushleyev et.al (2011)

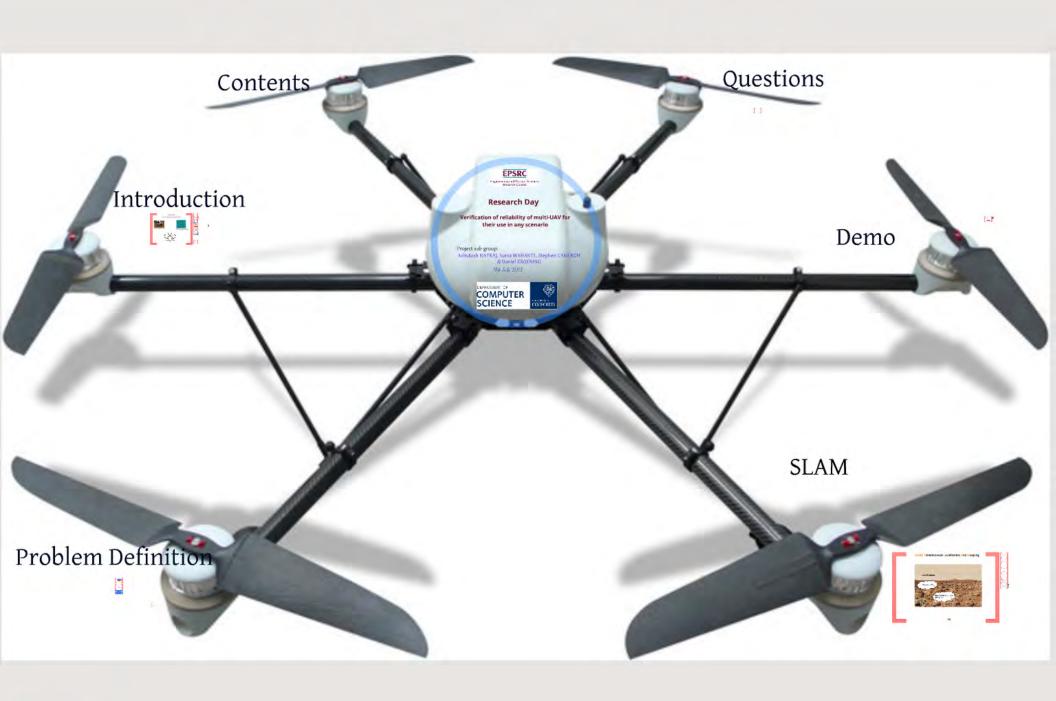
## **Computer Vision For UAVs**



Interest: Computer Vision for UAVs, but why?

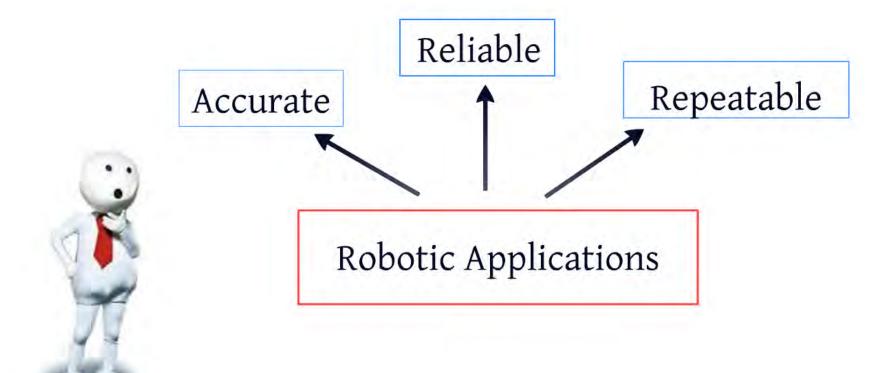
What is it? Visual processing on images captured from camera on board the UAVs.







# **Problem Definition**



But, what and how do we verify it?



#### **Problem Definition**

Uncertainty from sensor measurement

Dynamic memory allocation to cause read write conflicts

#### Errors can occur due to:

Error in positioning due to drift despite the verification of the code implementation and functioning.

# **Verification -Lends a Helping Hand**

Hardware level: Migration conflicts from intel core to ARM core

Low level: Memory allocation, read/write conflicts

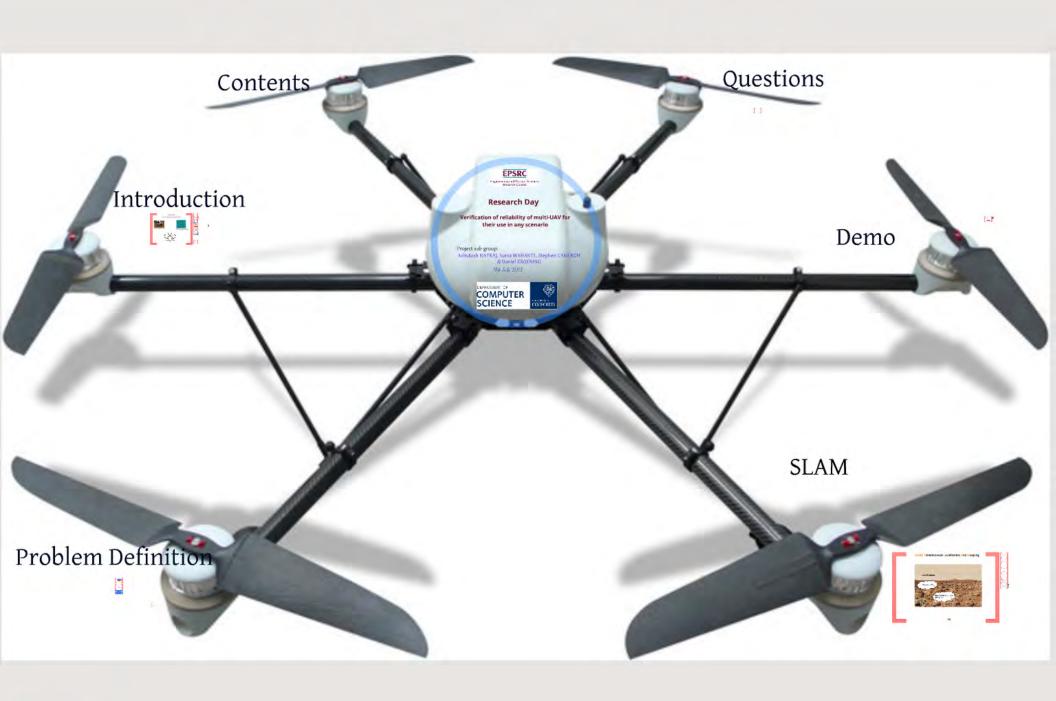
### What needs to be verified:

## High Level:

Algorithm Implementation

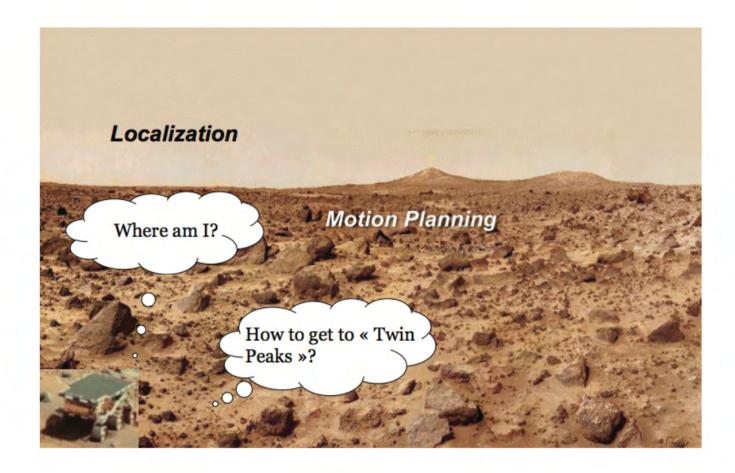
Some more yet to be decided and added in due course of time

Motor malfunction, heating and fire, Loss of sensor physical connection.





# **SLAM:** Simultaneous Localization And Mapping

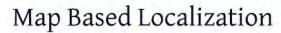


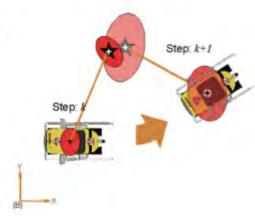
### **Localization Methods**

Dead Reckoning

Step: k

Uncertainty grows without bound and the robot is lost





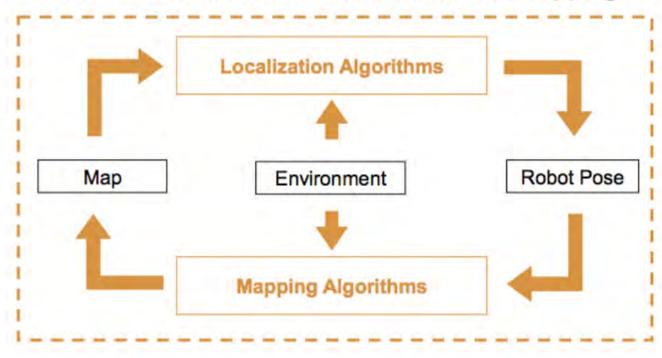
A conventional method for map building is incremental mapping.

Position Reference given by Dead Reckoning

Bad Maps => Poor Localization

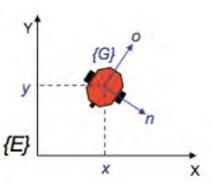
# SLAM: Simultaneous Localization And Mapping

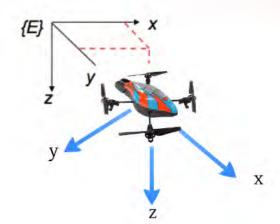
### **SLAM: Simultaneous Localization And Mapping**



#### **SLAM** Problem Definition

Pose: 2D:  $E(x, y, \theta)^T$ 3D:  $E(x, y, z, \phi, \theta, \psi)^T$ 





#### **Environment:**

Static: Only robot pose changes

Dynamic: Robot as well as the pose of other entities change

#### Localization:

Passive: Localization module only observes

Active: Robot is guided in a way that minimizes the localization error.

#### Given:

- Map of the environment
- Sequence of sensor measurements

#### Wanted:

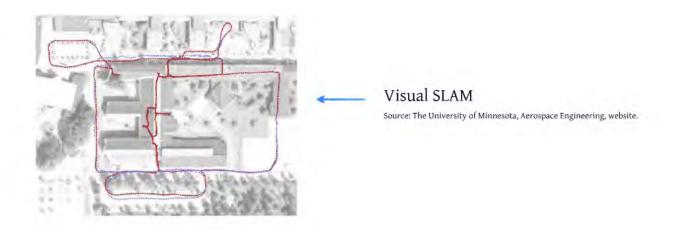
- Estimate of the UAV's position

#### Problem Classes:

- Position tracking (initial pose known)
- Global localization (Initial pose unknown)
- Kidnapped robot problem (recovery)



### Visual SLAM:



Visual SLAM as an important tool for localization as presented by Caballero et.al (2009) low cost, reliable, fast and hassle free.

Visual SLAM from partial structured environment by Artieda et.al (2009) relating features of objects tracked to their distance from UAV.

Visual SLAM from mosaics of images by Bailey et.al (2006)

### **Scenarios**

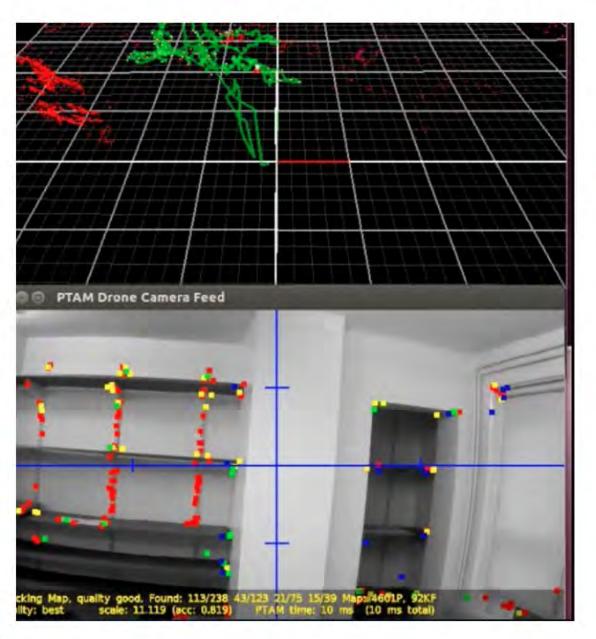
### **Scenarios:**

- 2D Environment: Hallway with 3 doorways.
- 3D Environment: The mapping and localization of UAV

Explained with Demo



# **UAV** - Localisation



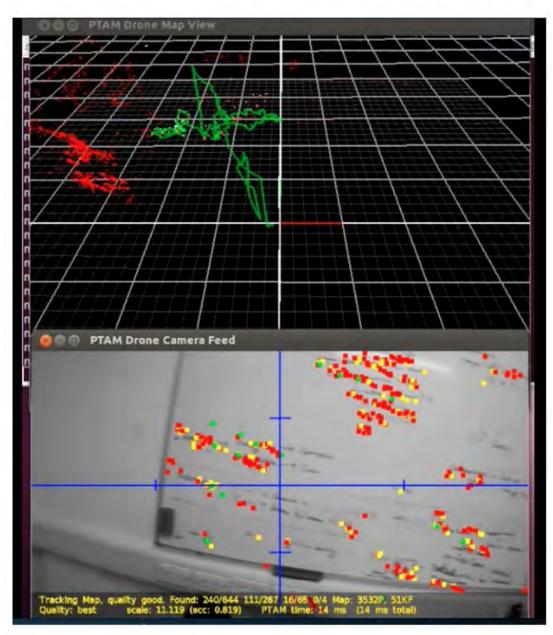
Green trajectory path - Localisation of the UAV

Red feature points – Generation and update of the map

Visual feed from the camera



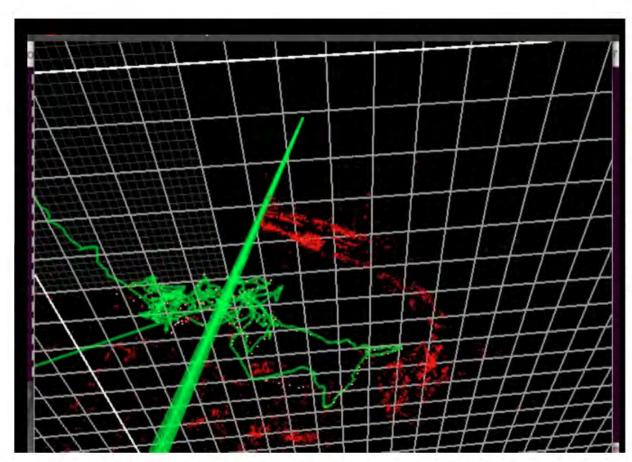
# **UAV** Localisation



View 2



# UAV – Map generation

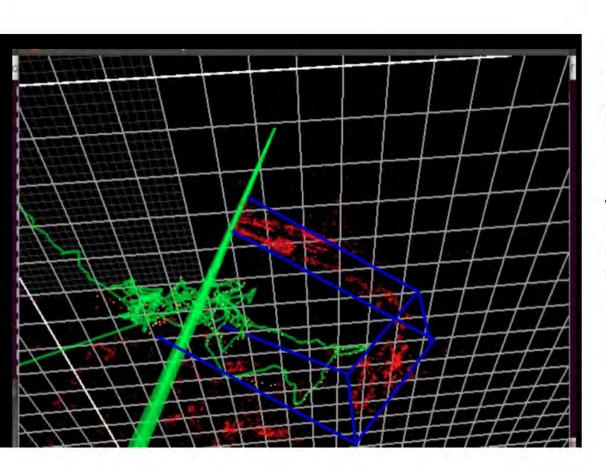


Map generated from Red feature points.

The map represents the 3 walls of the room .



# UAV – Map generation



Map generated from Red feature points.

The map represents the 3 actual walls of the room In Blue. The 4<sup>th</sup> is window.

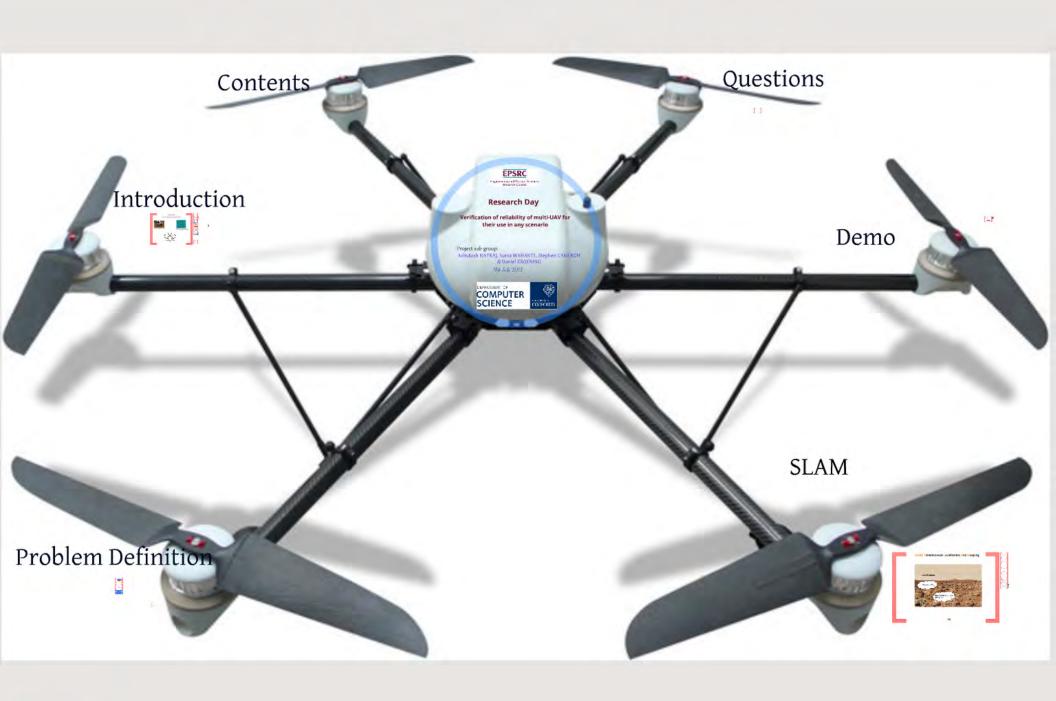


# Verification

Identified verification goals from implementation perspective:

- 1. Fault that arise from the motor malfunction.
- 2. Fault from drawing excessive current to the motors.
- Fault that arise from hardware failure loss of physical connection of the sensor (camera / IMU/ sonar) – loose connection.
- 4. Fault detection from lack of feature points due to sudden occlusion on the camera or in fairly uniform environment that lack in enough feature points.
- 5. Fault that arise from losing the wifi established







# **Experiment - Demo**

### Demo

UAV Localization from visual information obtained from the environment.

Verification of identified features















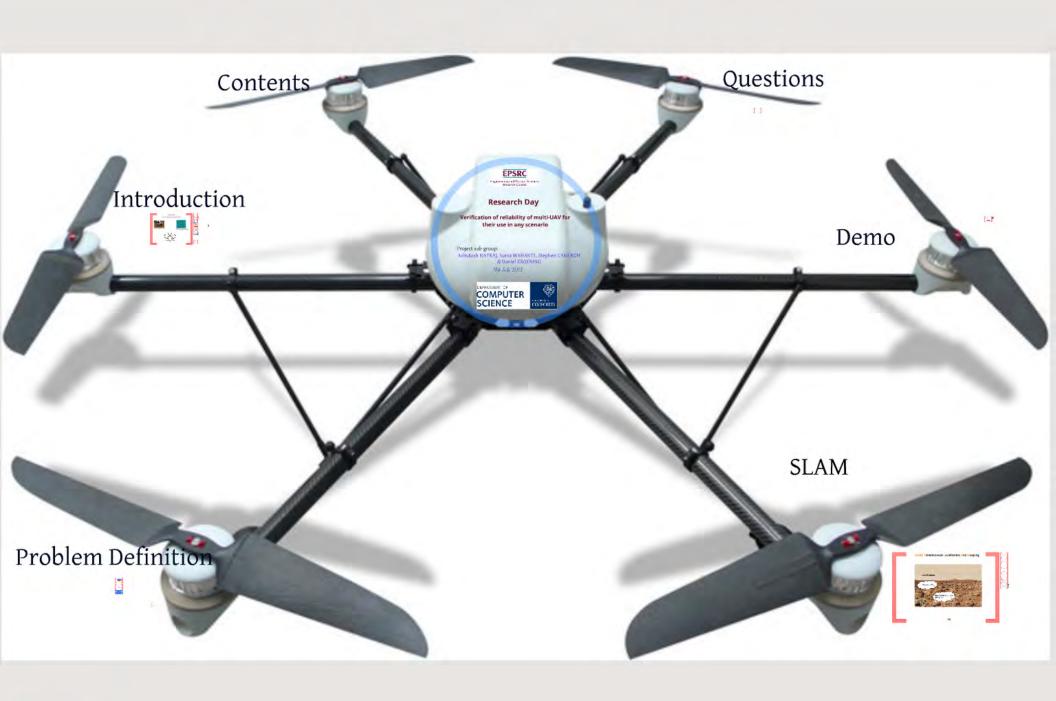
# **Previous Works**













# Questions

