

Localization and Mapping of Flying Robot in GPS-Denied Environments

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Abstract

Generally the search and rescue missions in the case of monitoring operations, natural and man-made disasters mainly rely on manned aerial vehicles. But these methods were highly risky for human pilot or rescuers because these included high navigation precision and long operation times, These unmanned aerial vehicles were used only in GPS prone areas based on an inbuilt map. These drawbacks are overcome in this proposed paper by designing a vision-controlled hexacopter which provides vision-controlled flying in GPS-denied environments. It consists of two stages: The onboard stage runs the visual odometry algorithm and the off-board stage runs the localization & mapping algorithm.

Keywords: Mapping, vision-controlled, odometry, Localization, Hexacopter

I. INTRODUCTION

The improvements in technology have led to the use of UAVs (unmanned aerial vehicles) over the last few decades. UAVs were used in several fields and replaced human pilots. UAVs offer promises of speed and access to regions that are inaccessible to ground robotic vehicles. They can be scaled down to small sizes and can operate in closed and confined environments. In search and rescue operations there are few constraints like time is critical and if there is any delay then it may result in human losses. The environmental conditions are also unfriendly that are difficult for the rescuers to easily reach. So by using UAVs a critical support is provided for search and rescue operations. UAVs are fast and can perform operations that are hard to be executed by human operators at low operating costs. UAVs can be deployed in an area of interest and report their collected information to a remote ground station or rescue team. This paper describes the technical challenges and the results achieved from hardware design and embedded programming to vision-based navigation and mapping, with an overview of how all the modules work [1].

II. RELATED WORKS

The occurrence of unpredicted natural and man-made disasters in the past decade has urged the search and rescue team to seek some effective equipment to enhance their efficiency. Usually the search and rescue operations mainly depend on search dogs or manned helicopters. But these included high navigation and operation times. So robots were introduced to overcome these drawbacks. The first real research on search and rescue robots was done after the Oklahoma City bombing in the year 1999. Though robots were not implemented, suggestions were made how robots can be implemented in such situations. Then in 2001 after the 9/11 WTC disaster robots of various sizes and capabilities were used. Carlos Marques *et.al* [2] described about RAPOSA an unmanned ground vehicle, that can operate in outdoor environments such as debris which resulted from the collapse of built structures. Then in 2006 came the unmanned aerial vehicles that surveyed the damaged environment caused by Hurricane Katrina. Pierre-Jean Bristeau *et.al* described Parrot A.R Dronas navigation and control technology based on a quadcopter design [3] that uses ATMEGA8L 8bit microcontroller. The various components needed by the quadcopter for accurate flying using vision as main sensor are a monocular SLAM system that provides localization and mapping, an extended Kalman filter for data fusion and state estimation and a PID controller to generate steering commands to reach the desired goal [4]. Sunantha Krishnan *et.al* gave an overview of Rescue Robot [5], that implements Viola-Jones face detection algorithm to detect survivors. Using real time video transmission the robot location is sent via Bluetooth and GPS receiver is placed on robot. The microcontroller controls the robot. Then using PIC16F877A controller a Mobile Rescue Robot was developed based on Wireless Sensor Network (WSN) [6]. A spy ball robot or a spy robot/camera in a ball shape which has the ability to move in any direction like tunnels and pipes can be used both for spying or inspection and rescue operations [7]. The spy ball is operated through a wireless connection and can transmit video through a cam mounted in the middle of the spy ball. But these

technologies possessed few drawbacks. The UGVs can be either wired or wireless. If they are wired then there is chance that it may get struck and suppose it is wireless then it is difficult to track their location ones they wander away from the operators site And in the case of UAVs they are mainly used in the GPS prone areas and are confined only to a specific location based on its inbuilt map. This paper is organised as follows, Section II consists of related works regarding the UAVs. Section III describes the proposed approach. Experimental results are shown in Section IV and Section V concludes the paper.

III. PROPOSED APPROACH

In this proposed work, a vision-controlled flying robot is designed for search and rescue operations in GPS-denied environments using ARM 11. Localization & mapping is provided with visual odometry and sonar data (data from ultrasonic sensor). This robot can fly by using an onboard camera and inertial measurement unit. These sensors do not require any external infrastructure so they can easily operate in unknown environments where GPS signals are low. These sensors provide a system that is capable for vision-aided navigation and mapping.

A. Block Diagram:

The block diagram shows the components used in this approach. The two main modules used here are local navigation module and global navigation module. Local navigation module represents the body of the robot and global navigation module is the ground station.

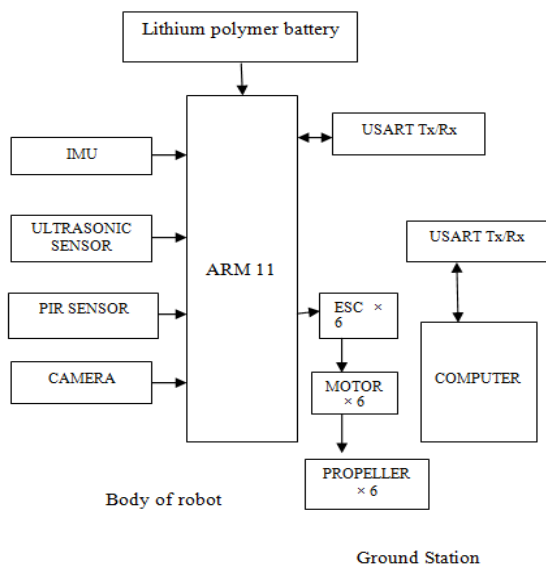


Fig. 1: Block Diagram of Proposed approach



Fig. 2: Body of the Robot

The robot is constructed using aluminium sheet and glass fiber rods. Initially two sheets of aluminium in hexagon shape with strong PCB material form the top and bottom layer of the hexacopter. Then to the six corners of this hexagon shaped PCB material, PA66+30GF thermoplastic rods are placed which are light weight and provide high rigidity. To the six ends of these rods brushless DC motors of 850KV are connected. Brushless DC motors are preferred since cost is low and has less friction. Hexacopter is preferred since stability is more [1]. Motor is a device that converts electrical energy into mechanical movement. Now the shaft of each motor is connected to a 10x4.5 propeller. ESC (electronic speed controller) of 30A is used to control the speed of the motor. The controller used here is ARM1176JZFS. The IMU (inertial measurement unit) used here is MPU 6050 i.e. accelerometer and gyroscope. The accelerometer provides stability and position and the gyroscope provides the angular velocity. Ultrasonic sensor is used to detect obstacles. The PIR sensor is used to detect lives, The data is transmitted and received using Zigbee. Zigbee S2 provides a range of upto 1 Km. The camera used is A/V camera of 5 Megapixel. The power supply is provided with the help of lithium polymer battery of around 2300 mAh. The hexacopter is then vision-controlled using visual odometry. Then localization and mapping can be done using camera and data obtained from ultrasonic sensor.

B. Visual Odometry:

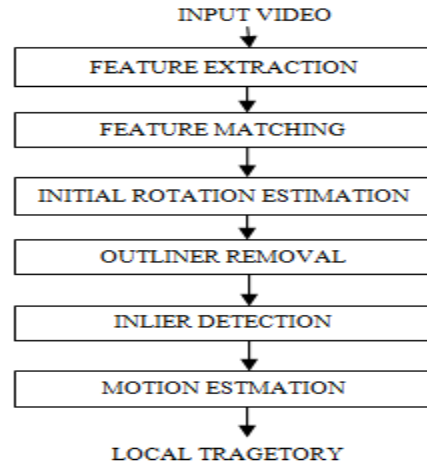


Fig. 3: Steps in Visual Odometry

Visual odometry is the method of estimating the motion of an agent based on the input obtained from a camera attached to it. Motion can be obtained based on the variation in the poses of the images captured from camera. Its similar to wheel odometry where motion is determined based on the turns in the wheels [9],[10].Steps in Visual odometry are:

- 1) Feature extraction: There are two methods to find feature points :a) Find features in one image and track them in another using correlation. This is used for nearby viewpoints. b) To independently detect the features in all the images and match them based on similarity metrics. This is used for far by viewpoints. The relative motion can be computed using appearance or feature based method. Appearance based method use the intensity information of all the pixels in the two input images. Feature based methods only use salient features extracted across the images.VO implements feature based approach.During feature detection image is searched for some key point's . Corners and blobs are few features that have to be detected. Corner is point of intersection of two edges and blob is pattern that differs in intensity and colours. Features are extracted using FAST feature detector where threshold is set in such a way that it extracts all features.
- 2) Feature matching: Each feature is assigned a descriptor of intensity values 9x9 pixels .Feature matching includes comparing all feature descriptors in two images. Features are matched using mutual consistency where by using hamming distance the sum of absolute difference is calculated. If the absolute difference is a small value then it is a perfect match.
- 3) Initial rotation estimation: Here minimizing of the sum of squared pixel error is done between the down sampled version of current and previous frames [8],[9].
- 4) Outliner removal: Matched points are contaminated by outliers which are wrong data associates. The causes of outliers are image noise, occlusions, blur .etc. For camera motion to be estimated accurately outliers has to be removed. This is the task of Random Sample Consensus (RANSAC)[11].This method performs model based hypothesis on a set of data points and compares it with other data points. The hypothesis that gives highest consensus is the solution.
- 5) Inlier Detection: Inlier points to the hypothesis are computed by point to equipolar line distance. But better than this method is the directional error method. This method measures the angle between a ray of image feature and equipolar plane [10].Finally the motion is computed after comparing a new image with reference image.

C. Localization & Mapping:

Determining the robots position when limited sensor information is available is a key challenge in robotics. Generally the robots position is determined using GPS or beacons. But these methods include expensive hardware cost and processing power. So in this paper localization & map construction can be done using camera and ultrasonic sensor data. Here there is no prior knowledge of environment and map is autonomously constructed by the robot. There are two algorithms [12]:

Positioning algorithm: To determine the robots position initially the feasible pose has to be determined where the expected view should match the data obtained from camera and ultrasonic sensor. Then it selects the best feasible pose. But to determine this pose mainly there should be information about the robots orientation which can be obtained from gyro or odometry. To evaluate this pose a map is needed, So here a 2-dmensional grid is used to provide a map of robots environment. This grid map consists of a matrix of cells and the values indicate whether the cell is empty or not. Using its sensors the robot determines the range vectors i.e. the distance from detected objects and compares against all occupied cells in grid. If a range vector is overlaid on a grid without interference by occupied cells, it indicates a feasible pose. Now certainty value in that grid indicates the likelihood that robot is located at that position. So when feasible pose is identified, the certainty value of corresponding cell is incremented. After all poses are obtained the grid cell with highest certainty value is selected as the robots present position.

Occupancy algorithm: This algorithm creates a map of the environment by integrating data collected over time. As the robot explores the environment, data from the sweeps is combined with information about the robots location to update the occupancy values for grid map. The probability of occupancy is given by Bayes rule. The occupancy value ranges from 0 to 1 and 0.5 if the cell is undecided or unexplored. Based on thresholds it can be determined if the cell is occupied, empty or unexplored [12]. The energy radiated by the sensor is in the form of a cone which has half beam-width angle of 12.5 degrees. In figure 3, it is shown that the probability that a cell is occupied along the arc is $P_{occ}(max)$ and it drops exponentially to $P_{occ}(min)$. But there are sometimes chances of occurrence of error. Dead reckoning algorithm is used to make the previous readings zero in order to prevent any errors.

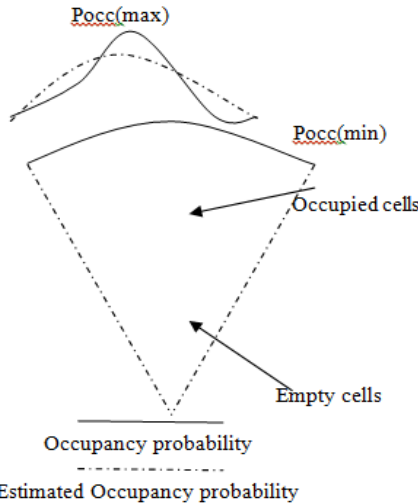


Fig. 4: Mapping using camera and sonar data

IV. EXPERIMENTAL RESULTS

The program is simulated in MATLAB. With the help of Zigbee the information is transmitted from body of robot to base station and vice versa. A 180-degree sweep is represented by red arrow. If there is any obstacle it is shown in red colour. The number of humans that are present in an unknown location is viewed using Terminal software.

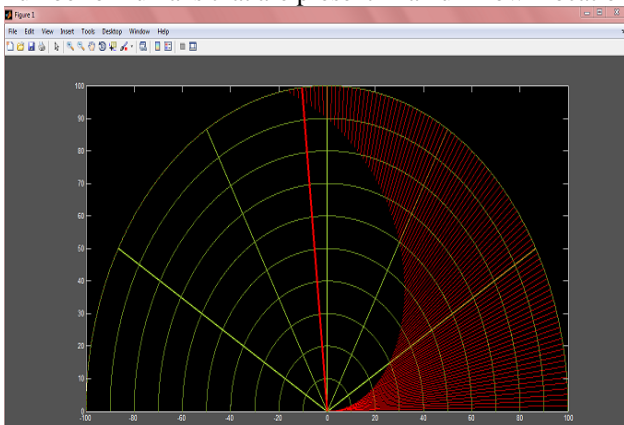


Fig. 5: Output viewed in matlab

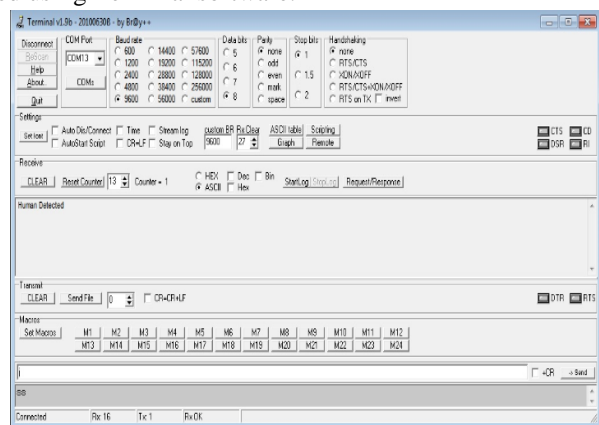


Fig. 6: Life detection viewed in Terminal software

V. CONCLUSION

This paper proposes the design of a flying robot that implements localization and mapping in GPS-denied environments. Visual odometry algorithm runs onboard and localization and mapping algorithm runs off-board. By using PIR sensors lives are detected in unknown locations.

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