

ÇANKAYA UNIVERSITY FACULTY OF ENGINEERING COMPUTER ENGINEERING DEPARTMENT

Project Report

CENG 407

Innovative System Design and Development I

Project ID: 202119 Autonomous Drone Control

Ahmet Çetin TÜRKYENER 201711066

Songül Meryem ÖZBİLEN 201711048

Elif Yağmur ERATALAY 201811028

Advisor: Ahmet COŞAR

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ABSTRACT

In recent years, drone technology shows itself in many areas. For example, it is no longer impossible to see drones flying in the air in the world of journalism and show, cargo distribution, agriculture, and emergency situations. Drones are unmanned aerial vehicles. It consists of components such as propeller, engine, body and flight control board. We will also automate the drone by using artificial intelligence and image processing technologies. Thus, it will be able to detect previously defined obstacles and reach the target. At the same time, we will ensure that when a break occurs, the situation is resolved promptly with the least cost, security issues are kept to a minimum, and the ability to react quickly to obstacles. In this literature review, we mention about the projects that have been done before on this subject, how we will do our own project, the problems that we may encounter and their solutions.

Keywords: Drone, Artificial Intelligence and Image Processing

ÖZET

Son yıllarda drone teknolojisi birçok alanda kendini göstermektedir. Örneğin gazetecilik ve televizyon dünyasında, kargo dağıtımında, tarımda, acil yardım durumlarında havada giden droneları görmek artık imkansız değil. Dronelar insansız hava araçlarıdır. Drone; pervane, motor, gövde ve uçuş kontrol kartı gibi parçaların birleştirilmiş halidir. Biz de yapay zeka ve görüntü işleme teknolojilerini kullanarak dronun otomasyonunu sağlayacağız. Böylece daha önceden tanımlanan engelleri algılayıp hedefe ulaşabilecek. Aynı zamanda, bir kırılma olduğunda en az maaliyetle hemen durumun çözülmesini, güvenlik sorunlarının minimumda tutulmasını ve engellere hızlı tepki verebilmesini sağlayacağız. Bu literatür araştırmamızda bu konu hakkında daha önceden yapılmış olan projelerden, kendi projemizi nasıl yapacağımızdan, karşımıza çıkabilecek problemlerden ve onların çözümlerinden bahsediyoruz.

Anahtar Kelimeler: Drone, Yapay Zeka ve Görüntü İşleme

1.LITERATURE SEARCH

1.1. Introduction

Drones are known as Unmanned Aerial Vehicles (UAV) with the name used in the technological field [1]. Drones are robots that can be controlled by remote control or fly automatically under the control of various software added to their embedded systems. Drones, which were used only to provide security in the military field in the past, are now used in many various and useful areas such as transportation, camera shots and fire extinguishing. Drones can perform tasks assigned to them, such as taking off, landing, and flying from one place to another, through a remote controller. These controllers can communicate with the drone using radio waves [2].

The aim of this study is to develop obstacle recognition skills for unmanned aerial vehicles by using Image Processing and Artificial Intelligence methods and in this direction, enabling drones to find their position by detecting the objects around them on autopilot.

In our Literature Review Report, we have Work Part containing Drone Preparation, Image Processing with Deep Neural Network Model and Ensuring that the software created with Image Processing and Deep Artificial Neural Network Model communicates with Pixhawk to automatically fly from desired locations. It also includes some of the problems that we may encounter during the project and their solutions. Finally, there is the conclusion part, where we evaluate the results of all our studies.

1.2. Work Part

1.2.1. Drone Preparation

It is the preparation phase of the test drone. Based on our research, we chose a drone that will be multicopter style [3]. In order to carry out the tests easily, to repair the drone at an affordable cost in the massacres, to make it ready for flight again, and to reduce the security problems, it is considered to build a small, 20-30 cm diameter, quadcopter-style drone with 5-7 inch propellers. We are planning to build a quadcopter-style drone [4], which is the system that gives commands to the motor that provide the balance of the flying device by processing the information coming from the gyro and these sensors through the software running on its processor. It is planned that the engines will be selected strong and the payload of the drone up to 500 grams. It is thought to have a mini camera on the drone, a mini computer for image processing and artificial intelligence. The minicomputer is planned to be Jetson Xavier [5], which will benefit us in

artificial intelligence coding and drone automation. According to the results of our research, it is planned to use open source code pixhawk autopilot as a 10 3A regulator autopilot on the drone to feed this computer and camera. This 10 3A regulator is used for all kinds of open source vehicles (multicopter, helicopter airplane, etc.), all kinds of multicopter types (tricopter, quadcopter, hexacopter, etc.) [6]. When a problem is encountered in the next tests, it is thought to have the option of turning off the engines with a remote command for safety reasons. The communication between Jetson and Pixhawk autopilot is planned to be done with UART communication ports, which is a communication protocol that provides communication between the peripheral units of the computers and microcontrollers on the Jetson and Pixhawk. It is considered to use 115,200 baud rate as the UART communication baud rate [7]. It is planned to create the structure of the UART communication system by having 8 bits of data as the UART communication protocol, selecting the parity bit, selecting 1 stop bit and selecting the timeout values and applying these values to both Jetson Xavier and Pixhawk autopilot. It is considered to use Linux Ubuntu operating system on Jetson Xavier as the operating system. The Pixhawk autopilot can communicate with standard 50 hertz motor drivers, but it is planned to increase motor driver communication to 400 hertz to make the drone respond faster. The images taken from the camera will be processed by Jetson using image processing and artificial intelligence and the guidance information obtained from this processing will be quickly transferred to the Pixhawk autopilot via the UART, and the Pixhawk autopilot will be able to fly with automatic responses at 400 hertz. The image processing model with the artificial neural network on our Jetson board is explained in the next phase.

1.2.2. Image Processing with Deep Neural Network Model

ANN, RNN and CNN models are the most commonly used artificial deep neural network models. The data types used by these models vary according to their structures. ANN uses Tabular Data Text Data types. CNN uses the Image Data type. RNN uses the Sequence data type.

Since the size of the datasets consisting of images is larger than the sizes of other datasets, it can be learned with a high-performance hardware and algorithm.

CNN is considered to be more powerful than ANN, RNN. Therefore, training a dataset consisting of images can be done by CNN with much higher precision. The filters used in CNN's Convolution Layer are also very similar to the feature extraction filters used in image processing algorithms, preventing the increase in the number of processes.

Whether the image size is large or small is not important for model training. The parameters that need to be trained are also related to the filter parameters, which strengthens our CNN selection.

ANN: Artificial Neural Network

RNN: Recurrent Neural Network

CNN: Convolutional Neural Network

| Type of NN | ANN | CNN | RNN |
|-------------------------------------|--|--|---|
| Type of Data | Tabular Data, Text Data | Image Data | Sequence data |
| Parameter Sharing | No | Yes | Yes |
| Fixed Length input | Yes | Yes | No |
| Recurrent | No | No | Yes |
| Connections | NO | INO | res |
| Vanishing and Exploding Gradient | Yes | Yes | Yes |
| Spatial Relationship | No | Yes | No |
| Performance | ANN is considered to be less powerful than CNN, RNN. | CNN is considered to be more powerful than ANN, RNN. | RNN includes less feature compatibility when compared to CNN. |
| Application | Facial recognition and Computer vision. | Facial recognition, text digitization and Natural language processing. | Text-to-speech conversions. |
| Main advantages | Having fault tolerance, Ability to work with incomplete knowledge. | High accuracy in image recognition problems, Weight sharing. | Remembers each and every information, Time series prediction. |
| Disadvantages | Hardware dependence, Unexplained behavior of the network. | Large training data needed, don't encode the position and orientation of object. | Gradient vanishing, exploding gradient. |

According to our research, first, a data set will be created with the images of the objects to be detected by the drone. This dataset creation process will be done with photographs of objects taken from different angles. The created dataset will not be a complete training set to be tested later, 80% of this dataset will be reserved for training and 20% for testing [8]. According to the results of our research, we planned to use CNN (Convolutional Neural Networks) artificial neural network on datasets for training. As an algorithm, we planned to use the YOLO (You Only Look Once) [9] algorithm. New training datasets will be processed with the YOLO algorithm using CNN. **YOLO (You Only Look Once):** It is an image processing algorithm using CNN (Convolutional Neural Networks). There are several reasons why we chose the YOLO algorithm. These are speed, accuracy and learning capacity for object detection and object recognition. YOLO [10] is generally superior to other algorithms in these aspects, so we plan to use this algorithm. The importance of these issues is outlined below.

Speed: In systems that will work in real time, objects must be detected very quickly. The drone is an agile platform that can move very fast, so speed is one of the most important issues for us. YOLO, which we will use as an algorithm, is 4-5 times faster than other object detection algorithms (such as RetinaNet-101, RetinaNet-50).

Accuracy: Accuracy is important as we are moving quickly with the drone live and YOLO can detect objects with very low error rate. After all, it is an algorithm that can detect objects both quickly and with high accuracy.

Learning capacity: The YOLO algorithm [11] is compatible with datasets expansion. Some other algorithms may suffer from poor prediction performance (especially in terms of accuracy) when there are many datasets. In the YOLO algorithm, a large number of objects can be defined quickly and accurately with extensible data sets.

YOLO's high object detection speed and prediction performance in the fast movements of the drone and its expandable learning capacity made us prefer this algorithm [12]. Our model, which will be formed as a result of our algorithm that will work with CNN on YOLO, will be run with our test data set, and our accuracy rate will be revealed according to the results obtained [13].

We will also include images of different objects in the test data, and in this way, the error rate of the trained model against incorrect images will show how the model will behave against different objects. Our accuracy rate will be determined as a result of the tests we perform with the test data.

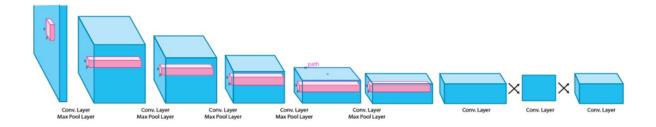


Figure 1: YOLO Architecture

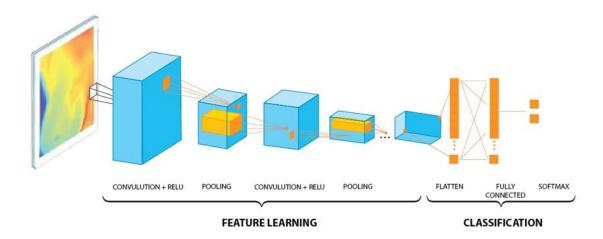


Figure 2: CNN Architecture

CNN (Convolutional Neural Networks)

First used in its modern form in 1990, CNN is a class of artificial neural networks most commonly used for image analysis in deep learning. Because CNN uses ReLU as activation, there can be a large number of data in the nerve and learning can take place without burden. CNN consists of basic foundation. These are Convolutional Layer - Pooling Layer - Fully Connected (FC) Layer.

Convolutional Layer

Our images in the training data have dimensions W \times H \times 3. W: width H: height 3: RGB (Red Green Blue) (Red Green Blue) value. W \times H in the images represents a matrix. Each cell of this matrix represents a pixel. This pixel has an RGB value. This value is between 0-

255. An N x N filter is determined according to the matrices of the images. The feature map of the image is created by sliding this filter over the image according to the Stride value (slip variable). This process is repeated many times with different filters. In this way, according to the attribute map, the object's color, corner, protrusion, etc. information is extracted.

ReLU

ReLU (Rectified Linear Unit) is a nonlinear function that operates as f(x) = max(0,x) [15]. So, if the relu function takes – it will output 0, if it takes + it will output that value itself. ReLU, whose main purpose is to get rid of negative values, has a very important position in CNNs. Nonlinear functions such as ReLU, tanh and sigmoid are used to prevent our model from learning negative values or not being able to grasp some features due to these negative values [14].

Max Pooling

Max Pooling is done to reduce the size of images with high dimensions without losing their properties [15]. Max Pooling is done by sliding over our W x H sized image with an N x N matrix. During this process, the highest value within a region of the size of our N x N max pooling matrix is determined as the new pixel value and other values are not used. Using a value directly belonging to the image without deriving a new value also prevents the image from being corrupted. Feature extraction takes place thanks to these processes.

Fully Connected Layer

Our matrix, which reaches up to the Fully Connected layer, is turned into rows by the Flattening process [2]. Values in rows are considered as Input Layer. Our neural network model has 3 stages as Input, Hidden and Output Layers. (Input Layer – Hidden Layer – Output Layer) Input Layer takes the incoming data in rows as input and transfers it to the hidden layers. Input layer node value * hidden layer node value gives the number of connections. The values coming to the Hidden Layer are connected to the Output Layer according to the output state we want. We will have outputs according to the number of objects we have determined in our project, plus the absence of any of these objects. We will have as many connections as hidden layer node value * output layer node value. The values we will end up with will be the object name or the absence of the object. As soon as the object is detected, our drone will react accordingly.

1.2.3. Ensuring that the software created with Image Processing and Deep Artificial Neural Network Model communicates with Pixhawk to automatically fly from desired locations

We are planning to develop the image processing and artificial neural network application, which we will develop based on the research we have done, in a structure that can process the image coming from the camera at 20 fps and make decisions at 20 hertz accordingly. In this way, it is aimed to process the frames as soon as they come from the camera and transfer the information accordingly. If the processor were on the ground, there would be several hundred milliseconds of delay due to RF (radio frequency) transmit and receive delays. But since the processor is on the drone, communication will take place via Jetson Xavier to UART, and from UART to autopilot at a baud rate of 115,200 via cables without RF (radio frequency) delay, so the drone will be able to react as quickly as possible. Considering that the drone can reach speeds of over 60 km/h, this minimum delay is very important in order to avoid problems. The system will make decisions with an accuracy of over 90 percent in each frame, and when the second frame arrives, it will work with 90 percent plus the remaining ninety percent of the ten percent, that is, 99 percent in total and with 3rd and 4th frames, this accuracy rate will increase much more [16]. We will test the system with many flight trials and we will be able to observe the accuracy rate. We will continue to work by updating the image processing and artificial neural network algorithms in case of potential problems and accelerating them if necessary [17].

1.3. CONCLUSION

As a result, the aim of this study is to develop possible obstacle models for drones by using Image Processing and Artificial Intelligence techniques and to enable the drones to pass these obstacles with autopilot and reach the target location. According to the results of our research, we will use CNN, Image Processing computer programming subbranches, PyCharm IDE, LabelIMG, Cloud GPU programs and YOLO reference algorithm method. We will use TensorFlow, Keras, OpenCV, NumPy libraries for this project. A low confidence score in object detection is seen as a potential problem. A possible solution for the related problem is to expand the dataset and add different angles to the visuals.

2.SOFTWARE REQUIREMENTS SPECIFICATION

2.1. Introduction

2.1.1. Goal

The goal of this project is to provide automatic flight-path control according to objects by implementing object recognition algorithms on the drone. Computational sensemaking of the objects seen with the unmanned aerial vehicle camera with image processing and artificial intelligence applications, processing this information and the information from the drone autopilot with a processor, creating new movement commands of the drone and transferring these commands to the autopilot are intended, which allows the drone to automatically move according to them without hitting the objects on the path.

2.1.2. Project Scope

Drones have become popular in recent years and have started to be used in almost every field that will benefit humanity in this regard. People from all over the world, students, academics, private and public sectors continue to launch new projects and new products related to drones. Despite being one of the fastest-growing areas of technology, drone technology has its own shortcomings. One of the most important of these is safe flight with object recognition. Drones cause the most accidents by crashing into objects while moving. They can crash into electricity poles, trees, hilly and mountainous places, buildings, and even more people, cars, etc. In addition, the delay in the use of drone technology near the ground (control patrols in factories, controls between buildings, close-up automatic drone commercials, etc.) is due to the inability of drones to fly safely in these areas due to crashes. Various technologies are being considered as solutions to these issues, and the implementation of object recognition algorithms on the drone that we are planning to do and the project of providing automatic path control according to objects is one of the most important steps that can be taken in this regard.

2.1.3 Advantages

With this project, drones will start to fly more safely and will get smarter. Accidents and crashes caused by drones hitting objects will begin to decrease, and human and animal injuries and deaths will begin to decrease. In this way, the use of drones in these areas will

increase, which is already expensive and therefore has reservations about its use in areas that can experience decimation. In addition, with drones getting smarter in the field of object vision, new technologies will be paved in this regard. For example, it can pave the way for hundreds of future drone technology issues such as automatic tracking of moving objects, dropping cargo on certain objects in a certain place, drones discharged from the automatic charging station by precision landing at the automatic charging station with object vision and continuing to the next patrols as automatic charging. As a result, accident incidents will decrease, new drone technologies will be paved, and its development and use will increase like any technology that has become safer.

2.1.4 Glossary

CNN: Convolutional Neural Networks

YOLO: You Only Look Once

ReLU: Rectified Linear Unit

GCU: Ground Control Unit [UAV ground control station (UAV – Unmanned aerial vehicles)]

2.1.5 Documental Overview

This document generally describes the safety of drones on objects on the path, the advantages to be gained if this security is provided, and how the technology will develop in this way. It gives information about autopilot software, flight hardware, communication hardware, mini-computer hardware and informs about image processing and artificial intelligence algorithms to be used on object vision, and the advantages of these algorithms over other algorithms in general.

2.2. General Description

2.2.1 Product Perspective (Product Features)

The project of Implementation of Object Recognition Algorithms on Drone and Providing Automatic Path Control According to Objects consists of hardware and software parts. As hardware, it basically consists of drone, drone autopilot (open-source autopilot), processor (jetson board) and camera. As software, autopilot software (open-source autopilot software), communication software, and image processing and artificial intelligence algorithm software that will work on the jetson card are combined and consist of drone control software, which is the decision-making mechanism.

2.2.2. Drone Preparation

We will use a small multicopter as a drone. The main reason for this is that drones are dangerous and expensive for testing. A multicopter with 10 inches and six propellers will do. We will build the drone completely ourselves. We will choose the engine, esc, propeller part, which is the thrust block, so that they are compatible with each other. As autopilot, we will choose an open-source autopilot, the autopilot we are considering is Pixhawk. We will have a telemetry for two-way communication. We will build the battery ourselves from lithium-ion cells. There will be a regulator that we can regulate 12V and 5V from the battery. We will design and manufacture the body with 3D programs according to the components we have chosen. On the front will be the camera zone. There will be a minicomputer place for image processing and artificial intelligence. Jetson Xavier was considered as a minicomputer.

2.2.3. Artificial Neural Network Model and Image Processing Methodology

A data set of objects that the drone can see will be created for image processing and artificial intelligence. This dataset creation process will be done with photographs of these objects taken from different angles. The drone needs to recognize these objects with great precision. For this, a large part (about 80%) of the photographic dataset will be used for training. The rest of the dataset will be used to test the correct functioning of the system. If necessary, we will enlarge this dataset until we reach the accuracy we wish.

Many systems are used for training and recognition algorithms. When we observed our system, we concluded that it should be a structure that will provide high accuracy solutions even when moving fast, especially since the drone moves fast.

As a result of our research, we chose the structure suitable for expanding the speed, accuracy and learning capacity when desired, Convolutional Neural Networks (CNN) artificial neural network on the training data sets and You Only Look Once (YOLO) algorithm as the algorithm. We decided to train our datasets with CNN and set up the algorithm with YOLO [18][19]. We found that the YOLO [20] algorithm is several times faster than other algorithms we researched, such as RetinaNet [21][22]. Speed is very important for us in the algorithm, as the drone can accelerate to 60 km/h. The algorithm should not compromise on accuracy while performing such fast processing, YOLO [23] can recognize objects with a very low error rate while performing fast processing. As we mentioned before, we will test our system with training datasets and expand our system with new datasets until we reach the desired performance. It gave us another reason to

choose YOLO [24] with extensible datasets and accurate prediction performance when multi-dataset.

CNN is widely used in deep learning to analyse images. CNN uses Rectified Linear Unit (ReLU) as its activation function. It is used for multi-layer deep learning models. Our visuals in the training data in the Convolutional Layer have dimensions W: width x H: height x 3: RGB (Red Green Blue). By performing MaxPooling, we can reduce the size of images with high dimensions without losing their properties. Thus, we will be able to make fast and accurate predictions. Our neural network model will have 3 stages (Input, hidden and output layer). We will have output as the number of objects we have determined from this model plus the number of cases where none of these objects are present.

2.2.4. Artificial Neural Network & Image Processing Algorithm Result - Drone Steering with Autopilot

Artificial neural network and image processing algorithm are planned to work on Jetson board. The routing information obtained with the outputs obtained because of this algorithm processed with the images taken from the drone camera will be transmitted to the autopilot from the Jetson card. It is planned to communicate with UART ports on Jetson and Pixhawk at 115.200 baud rate over UART. UART communication will be done over 8bit data, by selecting parity bit and 1 stop bit and timeout values. As the operating system, Linux Ubuntu operating system was chosen because it is more reliable and has less losses.

20 fps images coming from the camera will be able to be processed with our artificial neural network and image processing system. The whole system is on the drone and the system is independent from the ground control station. In this way, 2-way communication delays are prevented with telemetry, thus both the decrease in system performance and accidental crashes due to slowdown are prevented, and the task can be completed even in the presence of devices such as signal cutters in the environment. By seeing the system performance, system performance will be increased with extra data sets if necessary.

2.2.5 Restrictions

The biggest limitation of our project is the processing power, by implementing object recognition algorithms on the drone and providing automatic path control according to objects. Cameras have been minimized with the development of technology and they do not need much energy, but the processors that will take images from these cameras and process it and make artificial neural network analysis are not very small and their energy needs are higher. If these operations are carried out at the ground control station, we will

lose as much time as the signals coming and going from the drone, and since this loss of time will cause the drone to process and receive commands compared to the previous state of the drone, it is likely to experience troubles, accidents and incidents. In addition, when there are signal breakers in the environment, the connection between the ground control station and the drone may be cut off, so that the drone will be left without a command and crash. For all these reasons, we decided to keep the processor on the drone, despite the weight and processing power constraints.

2.2.6 Risks

The drone is an expensive system. In addition, it can accelerate very quickly, go fast and crash due to a technical problem or user error. As the drones get bigger and the carbon-fiber propeller, whose propellers are more efficient, the damage to the environment and people increases. Because of these risks, we decided to make the drone as small as possible for our project tests and use a plastic propeller instead of a carbon-fiber propeller.

2.3. System Requirements

2.3.1. External Interface Requirements

2.3.1.1. User Interface

The user will operate on the Windows operating system.

2.3.1.2. Hardware Interface

Required equipment is below:

- A gimbal camera with high resolution, UART communication, and high FPS.
- High performance minicomputer with UART communication, camera input, CUDA and CuDNN features
- Cables for communication
- HDMI cable for image transmission
- Drone body
- Propeller 4 Pcs.

- Engines 4 Pcs.
- Electronic Speed Controller
- Controls
- Autopilot
- Telemetry
- Battery
- GPS

2.3.1.3. Software Interface

We will develop an automatic flying drone as a prototype. We will use the open source Pixhawk Cube Orange autopilot for this drone and we will use the Mission Planner software interface for this autopilot. We will develop with Python programming language for autopilot software development, artificial intelligence software development and image processing software development, and the integrated software environment we will use for Python will be PyCharm. We will use TensorFlow and OpenCV libraries for Python. We will use the Linux Ubuntu operating system for Jetson Xavier.

2.3.1.4. Communication Interfaces

An extra communication interface not required.

2.3.2 Functional Requirements

2.3.2.1. Use Cases

Profile Management - Use Case:

Use Case:

- Start computer software
- Enter username / password
- Change settings
- Exit

Diagram:

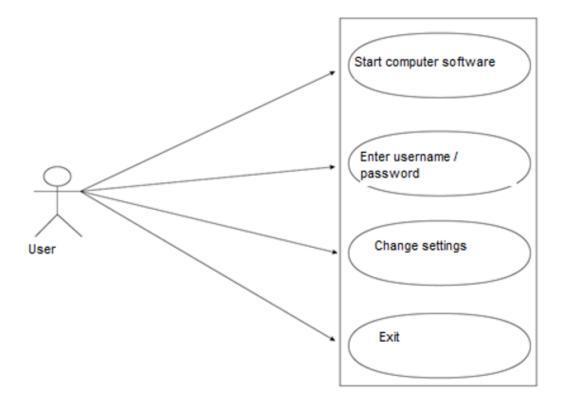


Figure 1: Profile Management - Use Case

Brief Explanation:

In our system, users cannot enter the system directly, users must be authorized. The use of drones is not legally open to everyone, people who have officially obtained the UAV0-UAV1 authorization certificate can use the drone. Users and passwords will be given to those who receive the authorization certificate and know how to use the system, so they will be able to change the settings of the system and use the system and log out whenever they would like.

Step-by-Step Explanation:

- 1. Anyone can start computer software.
- 2. Only authorized and qualified officials can log in to the system with a username and password.
- 3. The authorized user can change the system settings.

4. The user can log out of the system at any time when the usage is over.

Settings Menu - Use Case:

Use Case:

- Setting up a roadmap with waypoints
- Setting waypoints heights
- Adjusting drone speed
- Exit

Diagram:

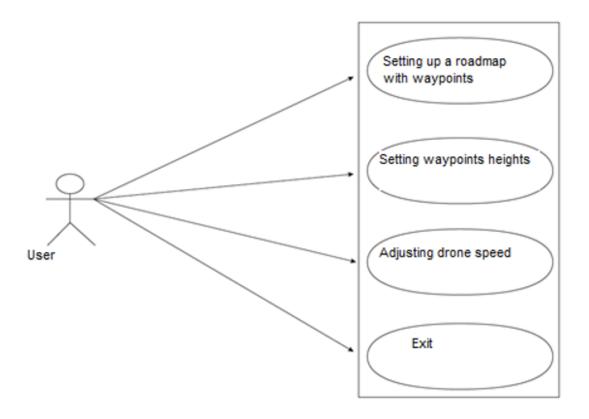


Figure 2: Settings Menu - Use Case

Brief Explanation:

With the settings, the user will be able to set the roadmap from google map. It will be able to determine from where and from which points the drone will pass. The drone will be able to adjust its speed. Height and speed adjustments can be made separately for the desired road parts, and the drone will not be active according to these new settings without sending these settings to the drone via telemetry. The user will be able to log out of the computer software at any time.

Step-by-step explanation:

- 1. The user sets up the roadmap by creating waypoints on the map.
- 2. User sets the height of waypoints; each waypoint can be at separate height.
- 3. User sets the drone speed. Standard drone can set speed or set speed separately for each path section.
- 4. The user can log out of the computer software at any time.

Detected Object Pass - Use Case

Use Case:

- Setting how many meters before objects change path
- Setting the transition of objects to the right, left, through or over
- Setting how many meters to pass objects
- Setting how many meters after passing objects to enter the defined path again
- Exit

Diagram:

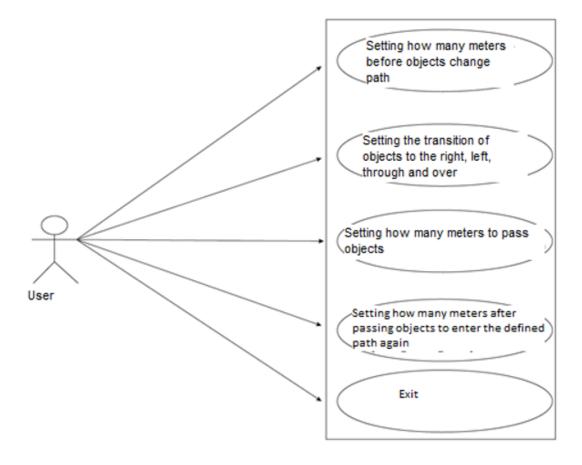


Figure 3: Detected Object Pass - Use Case

Brief Explanation:

Objects will be recognized in the system we train with artificial neural network and image processing algorithms. These settings are made on this menu for testing how close the drone should pass to objects and for drones of different sizes and different speeds.

Step-by-step Explanation:

- 1. According to the speed and size of the drone, how many meters before the objects should change the path, is adjusted.
- 2. According to the speed and size of the drone, it is set that it should pass from the right, left, inside or over the objects.

- 3. According to the speed and size of the drone, how many meters to pass the objects, is adjusted.
- 4. According to the speed and size of the drone, how many meters it will enter the defined path after passing the objects, is adjusted.
- 5. Exit from system after all settings.

Usage Menu - Use Case

Use Case:

- Send settings to drone
- Launch drone
- Send the new settings to the drone if necessary.
- Pause
- Land
- Take off
- Continue
- Finish the mission

Diagram:

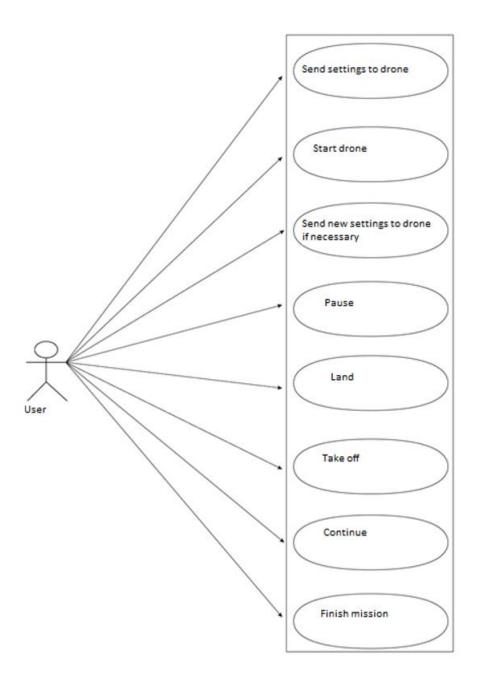


Figure 4: Usage Menu - Use Case

Brief Explanation:

The settings made for the drone are sent to the drone via telemetry and the drone starts, so that the drone automatically takes off and starts to go according to the roadmap. The drone travels completely automatically, so the user cannot make extra directions. When the drone encounters the objects in the data set while on the path, it applies the transition

protocol defined for that object from that object (from the right, left, under, inside). When it encounters an object that is not in the dataset, it applies the defined transition procedure for the objects that are not in the dataset. At any time, new settings are sent to the drone via telemetry, and as soon as these settings are sent, the new settings become active and the drone moves according to these settings. The user can pause the mission at any time so that the drone waits in the air, can land the drone at any time, take off and resume the mission at any time. The user can finish the task at any time, or if this command is not given until the drone finishes the task, the task will be completed automatically. When the mission is over, the drone will automatically land. If the user exits the computer software during the mission, the drone will finish the mission and land. After landing, GCU and drone are turned off.

Step-by-step Explanation:

- 1. The settings are sent to the drone via telemetry.
- 2. The command to start the drone mission is given from the computer.
- 3. The drone automatically takes off and starts to go according to the roadmap.
- 4. When the drone encounters an object in the dataset, it completes the previously defined transit protocol for that object.
- 5. When the drone encounters an object that is not in the dataset, it completes the previously defined transition protocol for the object that is not in this dataset.
- 6. The user can send the new settings to the drone via telemetry at any time, so that the new settings are active and the drone continues to fly with these settings.
- 7. The user can pause, land, take off at any time. Then he can resume the task.
- 8. The mission ends and the drone automatically lands.
- GCU and drone are turned off.

2.3.3. Performance Requirements

The requirements of the minicomputer that we will use for image processing and artificial neural network, are below:

- 1. 8 GB 128-bit LPDDR4x 59,7 GB/sec RAM or equivalent
- 2. 384 Core NVIDIA Volta™ GPU Equipped with 48 Tensor Cores or equivalent
- 3. 6-core NVIDIA Carmel Arm® v8.2 64-bit CPU or equivalent
- 4. 7-Way VLIW Image Processor or equivalent

2.3.4. Software System Features

Portability

The project will be able to work with open-source autopilots and autopilots with the appropriate SDK. Therefore, it can be transferred to most different drones on the market. It can be used not only for aircraft, but also for land and submarine vehicles.

Performance

- Each picture (frame) in the image taken from the camera will transmitted through the image processing software in 0.05 seconds.
- The received images will be transmitted through deep learning software within 0.04 seconds.
- It will be transferred to autopilot in 0.05 seconds.
- Image processing, deep learning, communication and autopilot software will run for a frame in a total of 0.14 seconds.

Availability

Our device will have two modes, semi-autonomous and fully autonomous. Thanks to these modes, the flight can be used completely automatically or with control assistance.

Adaptability

The image processing and artificial intelligence software and hardware used on the device are compatible to work with different autopilot hardware.

Scalability

The system has no scalability requirements.

2.3.5. Security Requirements

Since drones pose a danger to humans and animals by impact and system tests will be carried out, there should be no living being in the environment.

3.SOFTWARE DESIGN DOCUMENT

3.1. Introduction

3.1.1. Purpose

By making use of image processing and artificial intelligence software, it is aimed that the unmanned aerial vehicle detects certain objects and transmits information to the autopilot and completes its movement by avoiding the objects fully / semi-autonomously. In this way, the unmanned aerial vehicles, which are rapidly becoming widespread today, are fully automated according to the task, eliminating the human factor and minimizing the risks.

3.1.2. Scope

Unmanned aerial vehicles that can perform fully automatic missions, especially in military areas, are a very important need. Meeting this need will reduce the loss of people to almost zero in the military defence of countries. Drones have become popular in recent years and have started to be used in almost every field that will benefit humanity in this regard. People from all over the world, students, academics, private and public sectors continue to launch new projects and new products related to drones. Despite being one of the fastest-growing areas of technology, drone technology has its own shortcomings. One of the most important of these is safe flight with object recognition. Drones cause the most accidents by crashing into objects while moving. They can crash into electricity poles, trees, hilly and mountainous places, buildings, and even more people, cars, etc. In addition, the delay in the use of drone technology near the ground (control patrols in factories, controls between buildings, close-up automatic drone commercials, etc.) is due to the inability of drones to fly safely in these areas due to crashes. Various technologies are being considered as solutions to these issues, and the implementation of object recognition algorithms on the drone that we are planning to do and the project of providing automatic path control according to objects is one of the most important steps that can be taken in this regard.

3.1.3. Glossary

| Autopilot | Autopilot software is designed to limit the device, stabilize it, perform the task autonomously and device safety. | |
|-------------------------|---|--|
| Artificial Intelligence | Learning outcomes revealed by algorithms consisting of linear algebra methods used to calculate the pattern and trend in data sets prepared for a specific job | |
| Image Processing | By performing various matrix and mathematical calculations on the numerical units of visual elements such as photographs and/or videos, the desired information is obtained from the images, or the desired information is added to the images. | |
| CNN | Convolutional Neural Networks | |
| YOLO | You Only Look Once | |
| ReLU | Rectified Linear Unit | |
| GCU | Ground Control Unit [UAV ground control station (UAV – Unmanned aerial vehicles)] | |

3.1.4 Overview of Document

In this document, the integration process of artificial intelligence and image processing algorithms into an unmanned aerial vehicle and the process of gaining the ability to perform missions are explained. It is aimed to turn deep learning methods into software, to take matrix calculations on images as inputs in accordance with deep learning methods, and to calculate the outputs to be obtained by means of software. It is our ultimate goal that the device gains the ability to perform autonomous missions by transmitting the outputs to be obtained from image processing or artificial intelligence algorithms, which will work in harmony with each other, to the autopilot software.

3.1.5. Motivation

Artificial intelligence, which has rapidly increased in popularity in our country and in the world and adapted to people's lives very quickly, unmanned aerial vehicles that revolutionized the history of the world war, and our interest in image processing technology, which is an indispensable element of these two fields, played a key role in our participation in such multi-engineering work. At the same time, we, as senior students of Çankaya University Computer Engineering, who have adopted the ideal of carrying our country further than all the countries of the world in technological developments, are motivated to carry the name of our school and our country to the top.

Artificial intelligence technologies, which are at the base of the change and transformation in the world of computers and electronics, have started to be used in almost every sector. The most important revolution since the industrial revolution, where human power turned into machine power, is the development of artificial intelligence. Machines are developing not only in physical power but also in mental power. There are two most important factors needed for this development: 1. Data 2. Qualified people with high knowledge. As a country, it is aimed to extract these low-cost resources from our own essence and to integrate it into the development of artificial intelligence and our own products. If it works integrated with artificial intelligence, the thing that reaches high performances and will be indispensable in the near future is "image processing". The information about an image and the information that will be uploaded to those images by mathematical operations made on pixel basis over images such as photographs or videos are very exciting. And finally, autopilot technology, which is fully task-oriented and fully automatic, is the most basic element of unmanned vehicles. The combination of autopilot, artificial intelligence and image processing will pave the way for the emergence of a product that will minimize human loss. And we will be proud of the emergence of this product.

3.2. Software Architecture

In order to change the motion directions of the unmanned aerial vehicle, the photograph data of the selected objects will be read with the image processing software. Information about the pixel details of the photos will be taken as input to the artificial intelligence algorithm. As a result of the model formed, learning outcomes will be obtained. Objects will be detected according to the learning outcomes. The distance information of the detected object will be transmitted to the autopilot software via the communication protocol with the image processing algorithm. According to the information he received, the autopilot will provide the movements of the unmanned aerial vehicle.

- 1. The selected objects will be run on an independent device and will be learned by a deep learning algorithm.
- 2. Continuous pixel information of video images will be taken in real time.
- 3. Pixel information will be analysed in real time with an artificial intelligence algorithm that has been learned.
- 4. With the detection rate of 90% and higher by the artificial intelligence algorithm, the information of the detected object will be transmitted to the image processing algorithm.
- 5. The image processing algorithm will transmit the distance and direction information of the object to the autopilot as a result of the detection of the object.
- 6. The autopilot will maintain or change its direction according to the information it receives.

The system will operate with automatic mission-ready power-up. Against the risks that may occur, the user will be connected to the autopilot with a remote control and will enable the unmanned aerial vehicle to be directed.

3.2.1. Class Diagram

Autonomous Drone Autopilot Artificial Image Processing Intelligence readView() telemetry() readData() detectObject() manualControl() autoMission() calculation() sendData convertViewtoPixel() changeMode() threshold() getData() output()

Figure 1: Class Diagram

3.2.2. Decomposite Diagram

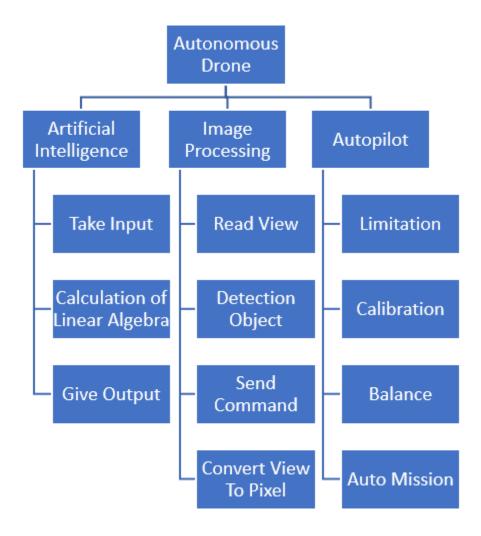


Figure 2: Decomposite Diagram

3.2.3. Data Flow Diagram

Level 0 of DFD

Mission: Mission is the process by which the drone detects certain objects and performs evasive maneuvers (with changes in speed and direction) and these occur fully autonomously.

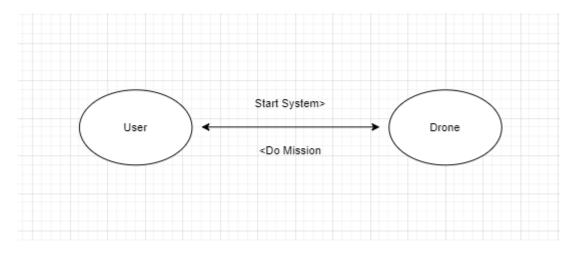


Figure 3: Level 0 of DFD

Level 1 of DFD

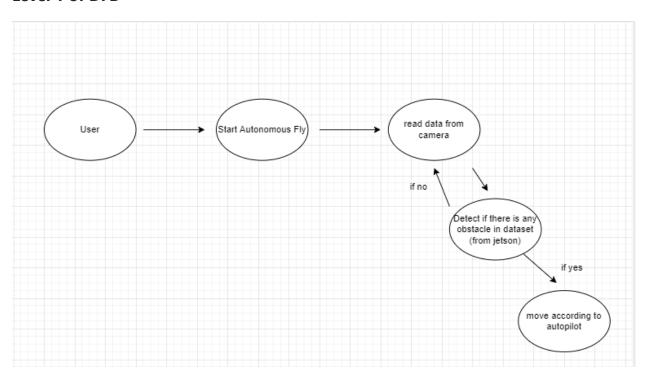


Figure 4: Level 1 of DFD

3.2.4. Activity Diagram

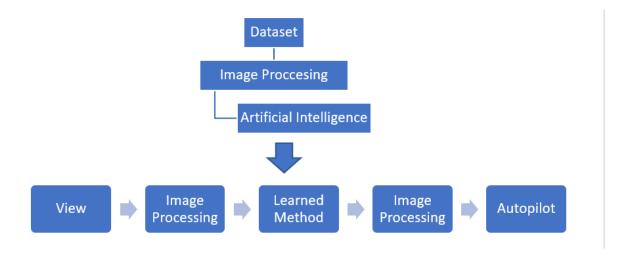


Figure 5: Activity Diagram

3.2.5. Sequence Diagram

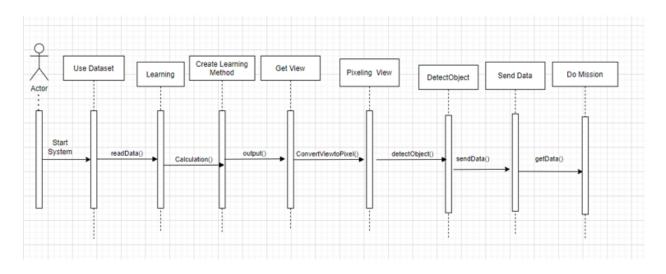


Figure 6: Sequence Diagram

3.4. Use Case Realizations

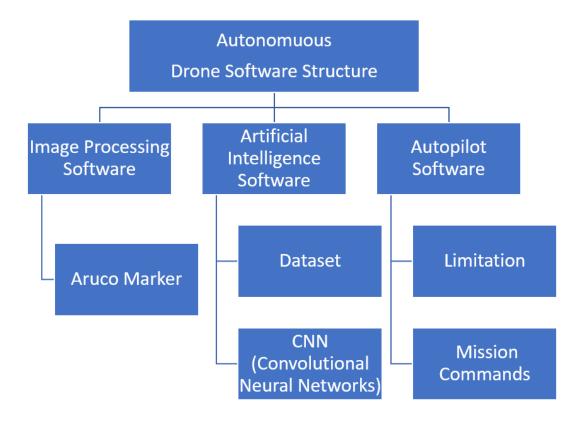


Figure 7: Software Structure of Autonomous Drone

3.4.1. Image Processing Software

It will be used to determine the location, distance and distinguishing features of the object. It will be ensured that the features of the objects are transmitted as input to the artificial intelligence algorithm at the pixel level. With some filters to be used, the distinguishing features on the image will be determined.

Aruco Marker

A standard size object that will be used in pixel cm conversion to determine the distance information of the object and to be calibrated for the first time.

3.4.2. Artificial Intelligence Software

It is the software that will be used in the formation of the learning algorithm and the learned model. This software uses the CNN structure, which is one of the deep neural

network models, and puts the inputs into various linear algebraic equations and gives learning output according to the result obtained.

Data Set

Data sets are at the heart of the Artificial Intelligence Algorithm's ability to provide learning. The more diverse the datasets contain, the higher the accuracy of the Al algorithm in learning outcomes. Having as many images as possible in the objects we selected while creating the data set will increase our learning percentage. If the learning rate is low, the data set can be expanded.

CNN (Convolutional Neural Networks)

Convolutional Neural Networks (CNN) is a class of artificial neural networks most commonly used for image analysis in deep learning. Because CNN uses ReLU as activation, there can be a large number of data in the nerve and learning can take place without burden.

3.4.3. Autopilot Software

It is software that enables unmanned vehicles to perform autonomous or semiautonomous tasks.

Limitation

Since device security is more important than anything else, developers make limits to secure the device with the help of autopilot. These limitations can be speed, incline, ascent, descent, etc.

Task Commands

According to the outputs from image processing and artificial intelligence software, information is transferred to the autopilot. The transmitted information is made meaningful on autopilot. Acceleration, deceleration, maneuvering, stopping and all other movements of our unmanned aerial vehicle according to the position of the object and the object information are transferred to the hardware with the autopilot software.



Figure 8: Mission Planner Moving Drone on Simulation Pixhawk

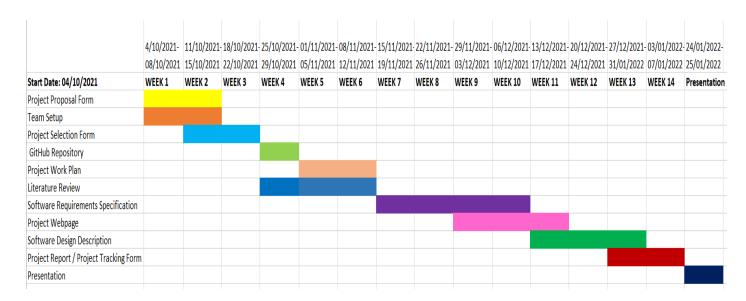


Figure 9: Mission Planner Do Mission

4.CONCLUSION & DISCUSSION

We have explained the documentation process of the Autonomous Drone Control project for our CENG407 course in detail in our report. The aim of this study is to develop possible obstacle models for drones by using Image Processing and Artificial Intelligence techniques and to enable the drones to pass these obstacles with autopilot and reach the target location. In Literature Search, we researched which algorithms, which software languages and libraries we should use for our Project. According to the results of our research, we will use CNN, Image Processing computer programming sub-branches, PyCharm IDE, LabelIMG, Cloud GPU programs and YOLO reference algorithm method. We will use TensorFlow, Keras, OpenCV, NumPy libraries for this project. In our SRS document, we created the requirements of our system. We have determined the use cases and their diagrams showing how the system will work. In our SDD document, we determined the architecture of our project's Image Processing, Artificial Intelligence and Autopilot software. In addition, we draw diagrams that show our software architecture and make our project more understandable.

5.PROJECT WORK PLAN



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