

Article

Generative Design, Simulation, and 3D Printing of the Quadcopter Drone Frame

Victor Andries and Sebastian-Marian Zaharia * 

Department of Manufacturing Engineering, Transilvania University of Brasov, 500036 Brasov, Romania;
victor.andries@student.unitbv.ro

* Correspondence: zaharia_sebastian@unitbv.ro

Abstract

The frame of a quadcopter drone is the most important component, as it supports all other systems and plays a vital structural role, supporting the stresses that appear during flight. The objective was to reduce the weight of the frame (18% reduction compared to the original version manufactured from carbon fibre), while maintaining structural integrity, by using an integrated strategy that includes optimizing the frame shape according to the components used, the stresses it must withstand, and considerations related to design for additive manufacturing. The optimization of the quadcopter drone frame was achieved using generative design and additive technologies (3D printing) and represents a cutting-edge approach in aerospace engineering, which allows for overcoming the limitations of traditional manufacturing methods. This study successfully completed all stages of the aeronautical product development cycle, from preliminary design, generative design, structural analysis, and 3D printing to assembly and functionality testing. The integration of generative design and 3D printing into the aeronautical product development cycle represents a complex and feasible challenge, with advantages in terms of efficiency, performance, and innovation capacity that fully justify the effort.

Keywords: generative design; FEA; quadcopter drone; 3D printing process



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1. Introduction

Topology optimization [1,2] and generative design [3] are modern techniques used in the life cycle of aerospace products with the aim of reducing manufacturing costs and weight, but also improving product performance. Topology optimization assists in the development of new products by providing a series of virtual iterations, calculating an optimal conceptual design based on a finite element model that satisfies a functional optimization objective, a set of geometric constraints, or a range of manufacturing requirements [4–6]. Topology optimization is a tool widely used by engineers, with applications in various fields: aerospace [7,8], automotive [9–11], architecture [12,13], and biomedical engineering [14,15].

Generative design is a computational design technique that can automatically study a design with various constraints defined by mechanical engineers. Generative models based on topology optimization use algorithms based on artificial intelligence and generate various topology models that cannot be represented by classical parametric design approaches [16,17]. Generative design is a modern technique with recent applications in engineering (aviation, manufacturing, automotive, and construction) that uses mathematical

principles and generative algorithms based on artificial intelligence, leading to a synergy between architecture, unique and innovative design, and information technology [18–20].

Generative design tools combine human creativity with artificial intelligence to generate numerous design models in parallel with cloud computing that meet predefined criteria based on the user's initial requirements, such as materials used, design environment and functional surfaces, performance requirements, and manufacturing constraints [18,20,21]. Generative design and additive technologies are closely linked, enabling CAD (Computer-Aided Design) models to be produced physically in the shortest possible time and at low cost. Additive manufacturing processes offer unlimited possibilities for creating innovative and customized products using various types of materials: composites, metals, and polymers.

Table 1 provides an overview of the key materials, advantages and disadvantages, and specific applications of the primary additive manufacturing processes.

Table 1. Comparative analysis of additive manufacturing processes [22–25].

Process Type	Materials	Advantages	Disadvantages	Applications
Material Extrusion (MEX)	Standard materials, Composite materials, High-strength polymers.	Low cost, easy to use, wide range of materials.	Low accuracy and resolution, Anisotropic properties, voids.	Prototypes, general use parts, design objects.
Direct Energy Deposition (DEP)	Metals (steel, titanium, nickel) and ceramics.	Ideal for repairing large parts, good mechanical properties, can use multiple materials.	Very high cost, requires extensive post-processing, limited precision.	Repair and re-manufacturing of the aerospace and automotive components.
Vat Photopolymerization (VP)	Liquid photopolymer resins.	Excellent precision and resolution, very smooth surfaces.	Brittle materials, requires post-processing (cleaning, UV post-curing).	Medical and dental industry, micro-fabrication and electronics.
Binder Jetting (BJ)	Powder (metals, sand, ceramics).	High production speed, ability to create coloured parts.	Lower mechanical strength (requires infiltration or sintering).	Functional metal parts, ceramic parts, medical products.
Material Jetting (MJ)	Photopolymers (liquid resins).	Very high precision, ability to combine materials and colours in one part.	High material costs, materials can be brittle.	Medical models, injection moulds, aerospace prototypes.
Powder Bed Fusion (PBF)	Polymer powder, metals and ceramics.	Functional parts, good mechanical properties.	High costs, requires post-processing (powder removal).	Aerospace and automotive components, medical implants.
Sheet Lamination (SHL)	Paper, plastic, or metal sheets.	Low material cost, relatively high speed.	Low precision, poor mechanical properties.	Aerospace, automotive and biomedical parts.

Additive manufacturing processes for drone components [26,27] have developed in the UAV industry, offering numerous advantages in terms of customization, topological optimization, material testing, and rapid prototyping [28–30]. A recent study [31] addresses a generative software tool that can help create aircraft or UAV (Unmanned Aerial Vehicles) models in CAD software systems based on parametric models, aircraft model databases, preliminary design methods, and technological and aerodynamic requirements.

Topological optimization and additive technologies have become essential tools in the design, manufacture, and testing of aircraft structures (UAVs, airplanes, helicopters, gliders, and satellites). These technologies make it possible to produce lighter, stiffer, and more efficient components, which is crucial in the aerospace industry [32–34]. A heavily researched area of topological optimization in the aerospace field is brackets. Brackets are vital mechanical components used to fix and secure the position of other parts or assemblies, with applications in the aerospace industry [35,36]. Topological optimization of brackets manufactured by selective laser melting (SLM) is an advanced engineering approach that combines the benefits of generative design with the production capabilities of SLM technology. The success of a topologically optimized part manufactured through selective laser melting (SLM) is evaluated using three essential categories of metrics: mechanical

and material properties, geometric accuracy and quality, and economic and process metrics. The SLM manufacturing process results in a customized bracket adapted to the specific requirements of each application and with low production costs [37–39].

Generative design is an innovative technique for obtaining and developing drone frames, representing an approach based on algorithms and artificial intelligence for the structural optimization of aviation parts [40]. In this way, drone models can have complex and efficient designs that would be difficult or impossible to achieve using traditional design methods [41]. An interesting approach to the use of generative design and additive manufacturing in the creation of a drone frame with the aim of improving the strength-to-weight and power-to-weight ratios for precision agricultural applications was taken by Balayan et al. [42]. Another study [43] explored the implementation of generative design in the manufacture of UAV frames, using the FFF (Fused Filament Fabrication) process and carbon fibre-reinforced nylon (PolyMide™ PA6-CF) filament. Following the 35 feasible solutions generated for the drone frame, as well as finite element analyses and experimental tests, minimum displacement (0.12 mm) was determined under a load of 1.4 kg, and the equivalent von Mises stress was 3.28 MPa, along with a significant weight reduction of 66.28% compared to the original model [43]. The frame of the DJI F450 quadcopter drone was redesigned using generative design in the Autodesk Fusion software system, and the FEA (Finite Elements Analysis) results indicated a lower displacement compared to the initial DJI F450 drone frame [44]. Generative design was used to generate a series of models for a cargo drone, and the final model generated had the following constraints: maximum structural efficiency, manufacturability using additive technologies, operational safety, and good functionality [45].

Based on the current stage, it can be stated that a series of studies have been conducted in the field of generative design applied to drones or components of drone structures. However, very few studies have completed the entire cycle of drone development, from design to final assembly/initial operational testing. The objective of this paper is to demonstrate the feasibility of manufacturing a quadcopter drone using additive manufacturing and generative design processes, integrating all stages of UAV system development, starting from preliminary design, generative design, finite element analysis, and 3D printing, through to component assembly and functional testing.

2. Preliminary Design and Selection of Drone Components

The preliminary design for a quadrotor drone covers the important aspects necessary to begin the process of designing, developing, and building the drone. Specific details are refined during the generative design stage, depending on the specific requirements and constraints of the drone. Figure 1 describes the main stages of the preliminary design of the drone.

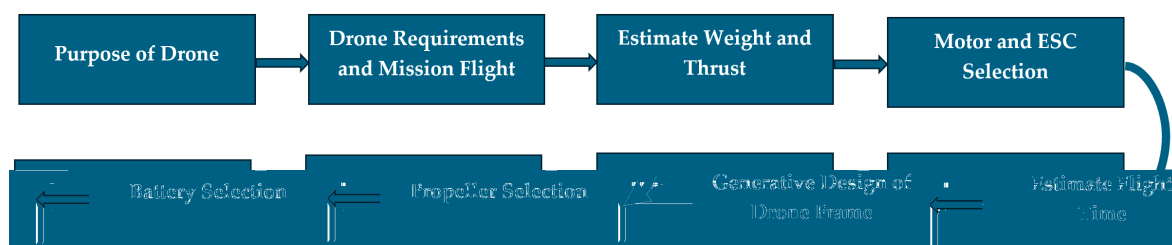


Figure 1. Flowchart of the preliminary design process of quadcopter drone [46].

The main goal of this study is to design a racing drone that is optimized for agility and high manoeuvrability, allowing it to perform the following missions: entertainment and sports; dynamic video capture using a GoPro camera; and improving piloting skills and

exploring flight limits. The drone developed in this paper is classified as C1 (according to the European Union Aviation Safety Agency—EASA), with a maximum take-off weight of 900 g, a maximum flight altitude of 120 m, and a maximum horizontal flight speed limited to 19 m/s [47]. The target mass from the generative design for the drone developed in this study is 750 g. The thrust-to-weight ratio (T/W) is a crucial factor in the flight performance of a drone, being the total thrust generated by the drone’s motors and propellers and the total weight of the drone. Since the main goal is to develop a racing drone, which requires a very high T/W ratio, usually between 5:1 and up to 10:1, for fast acceleration and sudden manoeuvres, the appropriate thrust-to-weight ratio is 9:1. To select the motor, the thrust per motor indicator was determined using Equation (1) [42].

$$\text{Maximum Thrust per Motor} = \frac{\text{Total weight of drone} \cdot 9}{4} = \frac{700 \text{ g} \cdot 9}{4} = 1687 \text{ g} \quad (1)$$

Based on the maximum motor thrust achieved, the HQProp FLUX motor (pull = 1676 g) was chosen, which does not exceed the thrust required for racing drones (1687 g), thus maintaining a high thrust-to-weight ratio (9.5:1). A stack that incorporates the FC (Flight Controller) and ESC (Electronic Speed Controllers) is a good option because it simplifies installation and reduces the risk of compatibility issues between the two components. Thus, the Mamba Stack with an MCU (Microcontroller Unit) was chosen, which can reach frequencies of up to 216 MHz. These MCUs are popular in-flight controllers because they offer a balance between performance and cost. Analysing the motor performance provided by the manufacturer, it was found that the tests were performed on standard 5-inch diameter propellers. Therefore, Foxeer Donut 5145 propellers, which have a diameter of 5 inches and two blades, which determine efficiency and high speed, were chosen for the model drone.

The selection of the drone battery was made in close connection with the characteristics of the motors, as they are the main consumers of energy, and their performance depends directly on the power supply. Considering the drone’s motors, a lithium-polymer battery with a capacity of 1550 mAh, voltage of 22.2 V, high discharge rate (120 C), and a good balance between power and low weight was used. A 22.2 V battery was chosen so that the motor-propeller assembly generates sufficient power, which is a good choice for a racing drone. The flight time of a racing drone is generally short, as they are designed for speed and agility, which require high energy consumption. The drone’s autonomy is calculated based on the motor and battery data, using Equation (2).

$$\text{Flight Time} = \frac{\text{Battery Capacity (Ah)}}{\text{Average Current Consumption (A)}} \cdot 60 \quad (2)$$

Table 2 shows the drone’s autonomy based on throttle level and current consumption.

Table 2. Determining the flight time of drone based on throttle level [48].

Throttle (%)	Battery Capacity (Ah)	Average Current Consumption (per Motor) [A]	Flight Time (min)
100	1.55	42.1	0.55
75		25.9	0.90
50		10.3	2.26
25		1.8	12.92

In a flight scenario, it is easy to calculate the flight time in mixed mode using the following flight time distribution: 100% throttle–5% of flight time; 75% throttle–5% of flight time; 50% throttle–20% of flight time; 25% throttle–70% of flight time. Using the percentages

established for this flight scenario and knowing the average current consumption, the autonomy of the drone in mixed mode is 3 min and 46 s. Of course, the autonomy of the drone varies depending on the flight mode and how much power is required to keep the drone in the air and perform manoeuvres. The last activity in Figure 1, the generative design of the drone frame, will be presented in detail in Section 3. Based on the preliminary design and the mission of the drone, the main components used in the development of the drone were established (Table 3).

Table 3. Components necessary for assembling the drone model.



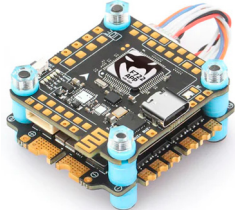






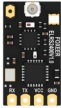
Drone Components	Characteristics	Image
Motor HQProp FLUX 2207 (Dancheng, China) [48]	Motor Diameter: 28.5 mm Output Shaft: 5 mm Motor KV: 1980 Voltage: 22 V Pull (Load current 42.1 A; Throttle 100%): 1676 g Weight: 34 g	
Propeller (Shenzhen, China) [49]	Size of Propeller: 5.1 inches Weight: 4.3 g; 2 × CW; 2 × CCW	
Flight Controller (FC) and Electronic Speed Controllers (ESC) MAMBA MK4 F722 APP (Shenzhen, China) [50]	Size: 38 mm × 40.5 mm × 8 mm Power: 3S–6S Lipo ESC Signal: 8 Set Frequency: 216 MHz Weight: 10 g	
Battery Tattu (Shenzhen, China) [51]	Capacity: 1550 mAh Voltage (V): 22.2 V Discharge Rate: 120 C Configuration: 6 Cells Weight: 253 g Dimensions: 76 mm × 38 mm × 47 mm	
FPV Video Transmitter Module Rush VTX Tank II (Changzhou, China) [52]	Input Voltage: 7–36 V DC Output Voltage: 5 V 1 A Channels: 48 CH Dimensions: 36 mm × 36 mm × 4.5 mm Weight: 6.8 g Power: PIT/25/200/500/800 mW	
FPV Caddx Ratel 2 Camera (Shenzhen, China) [53]	Image Sensor: 1/1.8 inch starlight sensor Resolution: 1200 TVL Dimensions: 19 mm × 19 mm × 20 mm Weight: 5.9 g	
RUSHFPV Cherry2 Antenna II (Changzhou, China) [54]	Bandwidth: 5600 MHz–6000 MHz Polarization: RHCP Radiation Efficiency: 99%	 MMCX-JW 83mm

Table 3. Cont.

Drone Components	Characteristics	Image
iFlight Analog FPV Goggles–DVR (Huzhou, China) [55]	Frequency: 40 CH 5.8 G Hz Size: 155 mm × 144 mm × 113 mm Screen size: 4.3 Inch Resolution: 800 × 480	
RadioMaster Boxer ELRS (Shenzhen, China) [56]	Frequency: 2.400 GHz–2.480 GHz Voltage Range: 6.6–8.4 V DC Channels: Max 16 channels Display: 128 × 64 Mono-chrome LCD display Weight: 532.5 g	
Foxeer ELRS 2.4 G Receiver LNA (Shenzhen, China) [57]	Work Frequency 2400 GHz~2500 GHz Input: 5 V Weight: 0.7 g Size: 10.5 mm × 19 mm	

3. Generative Design of the Quadcopter Drone Frame

For this study, the reference parameters were provided by the Mark4 HD drone [58] made of carbon fibre, weighing 163 g and measured as shown in Figure 2.



Figure 2. Reference drone frame [58].

The Autodesk Fusion 360 software system (Autodesk, 2023) was used for the generative design process of the drone frame. The main stages of the generative design process are described in Figure 3.

To establish the initial parameters of the generative design, the dimensions of the drone frame in Figure 2 were taken, and the dimensions of the new frame were defined as follows: motor layout at a diagonal distance of 240 mm; width of 184 mm (Figure 4a). Preserved geometry refers to those areas of the 3D model that cannot be modified by generative design. These preserved areas are divided into four categories (Figure 4b): through holes for mounting the GoPro camera mount and electronic components such as the flight controller, electronic speed controllers, and video transmitter; through holes for

mounting the electric motors; mounts and holes for the first-person view (FPV) camera; and rail-type supports (length 130 mm) for the battery, designed to allow the battery to be moved front to back, thereby changing the centre of gravity of the drone. The battery will be positioned above the drone electronics for easy access and quick installation and removal. Obstacle geometry is represented by those objects or shapes necessary to restrict the AI-assisted generative design software so that it cannot create material structures in those areas. In this study, these obstacle geometries were divided as follows: obstacles to the electronics placed in the central area of the drone frame structure; obstacles for the passage holes and for maintaining the shape of certain required geometries; obstacles created by motors and the space required for propellers; and other necessary obstacles created during the modelling/generation process of the drone frame.

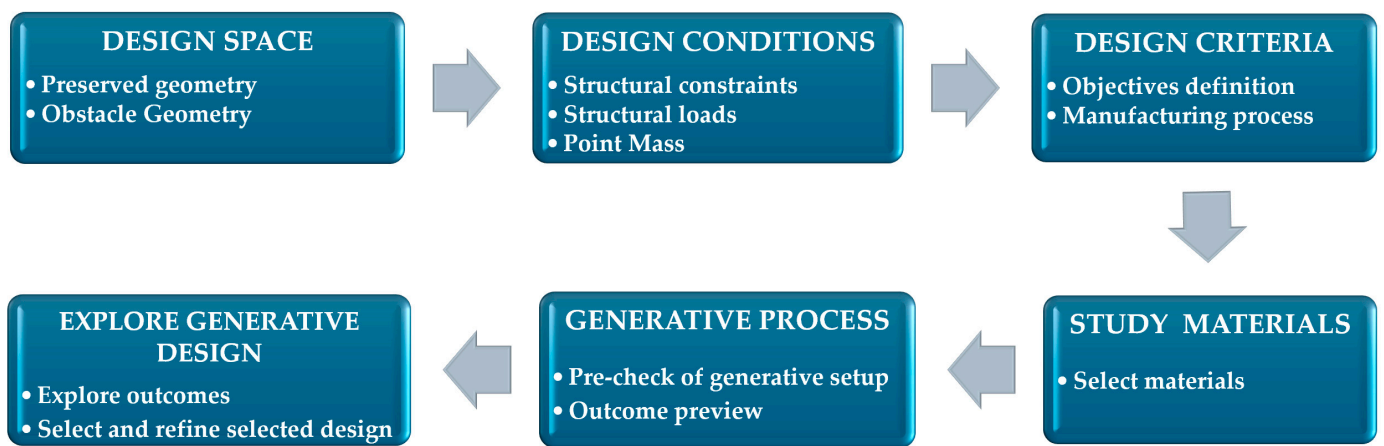


Figure 3. Workflow specific to the generative design process using the Autodesk Fusion 360 software system [42,59,60].

At the rear of the frame is the space required for the video transmitter, with the central space reserved for the FC and ESC. At the front is the space between the two geometries reserved for mounting the FPV camera. Between them are the obstacles for the cable passage and access to components. Above them, supported by the four mounting slots, is the GoPro camera mount, and further back is the obstacle created for the battery, which is supported by the rails of the required geometry. The positioning of these obstacles is strategic, ensuring easy access for the subsequent assembly of components. The obstacles for the through holes can be seen in Figure 4c. These obstacles were created so that the generative design does not fill the holes needed to mount the motors, electronic components, GoPro mount, and FPV camera, as this would complicate the assembly stages and render the generated frame unusable. To create obstacles, the dimensions of the propeller and electric motor were taken into account in the rotor area of the drone. The drone will be powered by four HQProp FLUX 2207 motors [48], with a motor diameter of 28.3 mm and a length of 32.2 mm, and a Foxeer Donut 5145 toroidal propeller's diameter [49] is 130 mm. The geometry and final layout of the obstacles can be seen in Figure 4d. The obstacle created as a plate shape (Figure 4d) is located under the mounting locations of the motors and electronics in the central part of the drone. It was added after the first iterative generation because in the initial design the structure would have passed through the lower part, which would have made it impossible to position the drone on the ground or land in a stable and safe manner.

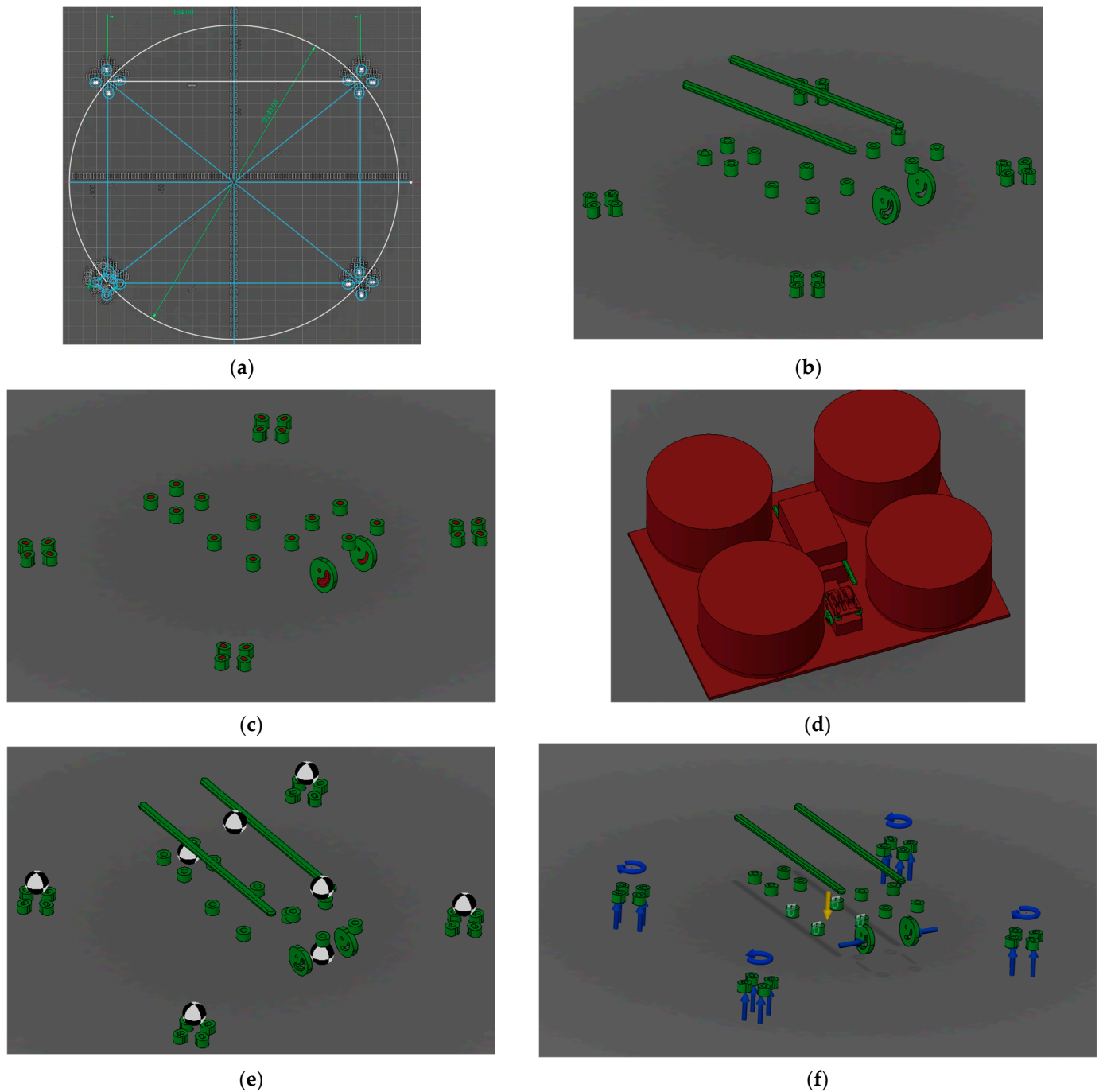


Figure 4. Stages of the generative design process applied to the drone frame: (a) initial sketch; (b) preserved geometry assembly; (c) obstacle clearance holes; (d) final obstacle geometry of drone frame; (e); arrangement of centres of gravity; (f) constraints and forces applied to the drone structure.

Defining parameters refers to all requirements, limitations, and tasks that were established before the actual start of the design generation, as well as the objectives proposed in this iterative study, for example, the minimum safety factor and target mass of the final frame. Using a safety factor of 2 for a drone frame manufactured using 3D printing is a prudent measure; a safety factor of 1.5 is commonly used in aerospace design. This higher safety factor contributes to increased structural safety of the drone, as it compensates for the disadvantages associated with material extrusion processes, such as material voids and anisotropy of the extruded material. It should be noted that these parameters are variable, and their values represent baseline values in model generation, with the software system

optimizing the final result. The parameters that formed the basis for the generative design of the drone frame are shown in Table 4.

Table 4. Main parameters established in the generative design process of the frame.

Parameters	Values and Purpose of Generative Design
Primary objective	Maximizing stiffness
Minimum target safety factor	2
Target mass of the drone frame	130 g
Manufacturing type	Additive
Material type for additive manufacturing	Tough PLA–Ultimaker

Figure 4e shows the centres of gravity of the components (Table 3) to be mounted on the drone frame, as follows: the motors are positioned centrally relative to the corresponding mounting holes, the electronic components and the GoPro camera are similarly positioned, and the battery has its centre of gravity positioned symmetrically between the support rails and centrally between the front and rear.

The last step before generating the design is to apply the constraints and demands that the frame must withstand (Figure 4f). The main constraint (fixed support) is on the four mounting areas of the FC and ESC, as that area must not deform during flight for the accelerometer and electronic gyroscope to function correctly, thus stabilizing the drone. The loads applied to the drone frame were as follows: thrust force of 20 N for each rotor (1676 g \cong 16.43 N–pull throttle 100% of the motor technical data multiplied by a safety factor [61,62] of 1.2); weight force–7.3 N (summing the mass of all components approximately as 750 g multiplied by gravitational acceleration); the moment per rotor 0.2 N·m was determined using Equation (3) [42]; the compression force of the FPV camera was 2 N.

$$\text{Rotor Moment} = \frac{60 \cdot \text{Maximum Current (A)}}{2\pi \cdot \text{KV}} \tag{3}$$

A comparative analysis (Table 5) of materials is essential for choosing the best option for a generatively designed drone frame, as the properties of each material (PLA, Tough PLA, and Acrylonitrile Butadiene Styrene—ABS) directly influence the performance, durability, and ease of manufacturing of the final part.

Table 5. Analysis of materials used for additive manufacturing of the drone frame [63–65].

Property	Material	Standard PLA	Tough PLA	ABS
Stiffness (Young’s Modulus)		Excellent (3.5 GPa)	Moderate (2.8 GPa)	Low (1.9 GPa)
Charpy impact strength		Very Low (3.9 kJ/m ²)	Good (8.9 kJ/m ²)	Excellent (14.2 kJ/m ²)
Hardness		Excellent 84 Shore D	Moderate 80 Shore D	Low 76 Shore D
Durability/Toughness		Brittle, snaps easily	Durable, can bend under stress	Very durable, bends under stress
Heat Resistance		Low (58.8 °C)	Low (58.3 °C)	Good (86.6 °C)
Ease of 3D Printing		Very Easy	Very Easy	Difficult
Warping		Minimal	Minimal	High
Typical Applications		Models, decorative parts, static prototypes	Functional parts, drone frames, mechanical components	Automotive parts, electronic housings, durable prototypes

After analysis of the three materials (Table 5) used in additive manufacturing, Tough PLA was the best choice for the prototype frame of the quadcopter drone. Tough PLA

offers the optimal balance between ease of 3D printing and mechanical performance, with no delamination or warping during the printing process. It also provides the rigidity necessary for stable flight, while offering the impact resistance needed to survive minor accidents, making it a more reliable option than standard PLA material, without the 3D printing challenges of ABS (warping, layer delamination, temperature sensitivity, and toxic emissions).

Obtaining solutions generated by generative design varies depending on the requirements and constraints imposed, with 13 different outcomes generated to observe the drone frames generated by the iteration method. Some of the outcomes have lower safety factors and obviously a minimization of the drone frame mass (Model 1—Figure 5a and Model 2—Figure 5b), while other outcomes have a defined target mass and the main goal of maximizing stiffness (Model 3—Figure 5c, Model 4—Figure 5d, Model 5—Figure 5e). Examples of outcomes generated by generative design based on the main parameters (Table 4) are described in Figure 5.

The following factors were considered when selecting the final drone frame design:

- Final weight—a lighter drone frame means greater flight autonomy and higher manoeuvrability;
- Structural strength—the drone frame must be sufficiently rigid so as not to deform under flight loads and have high structural performance to support the components (motors, battery, and flight controller) and withstand possible minor accidents. Also, the maximum stress appearing in the frame simulation must not exceed the maximum strength of the material from which the drone is made;
- Manufacturing method—generative design is often optimized for additive manufacturing. At this stage, the 3D printing method for the drone frame, areas with difficulties in 3D printing, and the creation of supports and their removal method were analysed and simulated (using the Ultimaker Cura 5.1.1 additive manufacturing preparation software system);
- Space and integration of drone components—the frame must provide adequate space and secure mounting points for all essential drone components (motors, propellers, battery, flight controller, and camera).

A multi-criteria analysis of the five most representative solutions for the drone frame (from Figure 5) is presented in Table 6. Thus, for the five models generated (Figure 5), a score was assigned based on their performance for each criterion, on a scale from 1 to 10 (1 = very poor, 10 = excellent).

Table 6. Multi-criteria analysis of the five drone frame models created using generative design.

Model	Mass (25%)	Stiffness (25%)	Safety Factor (20%)	Equivalent Stresses (15%)	Manufacturability (15%)	Total Score
Model 1	9	5	7	5	6	5.86
Model 2	8	7	7	5	7	6.70
Model 3	6	9	8	7	8	7.23
Model 4	7	9	8	7	8	7.79
Model 5	7	9	9	8	9	8.29

After studying the models performed by generative design, using a multi-criteria analysis (Table 6), the generative model with the highest final score (model 5—Figure 5e) was selected. Thus, a balanced solution was chosen that maximizes overall performance in relation to the design priorities and constraints, rather than excelling in a single criterion.

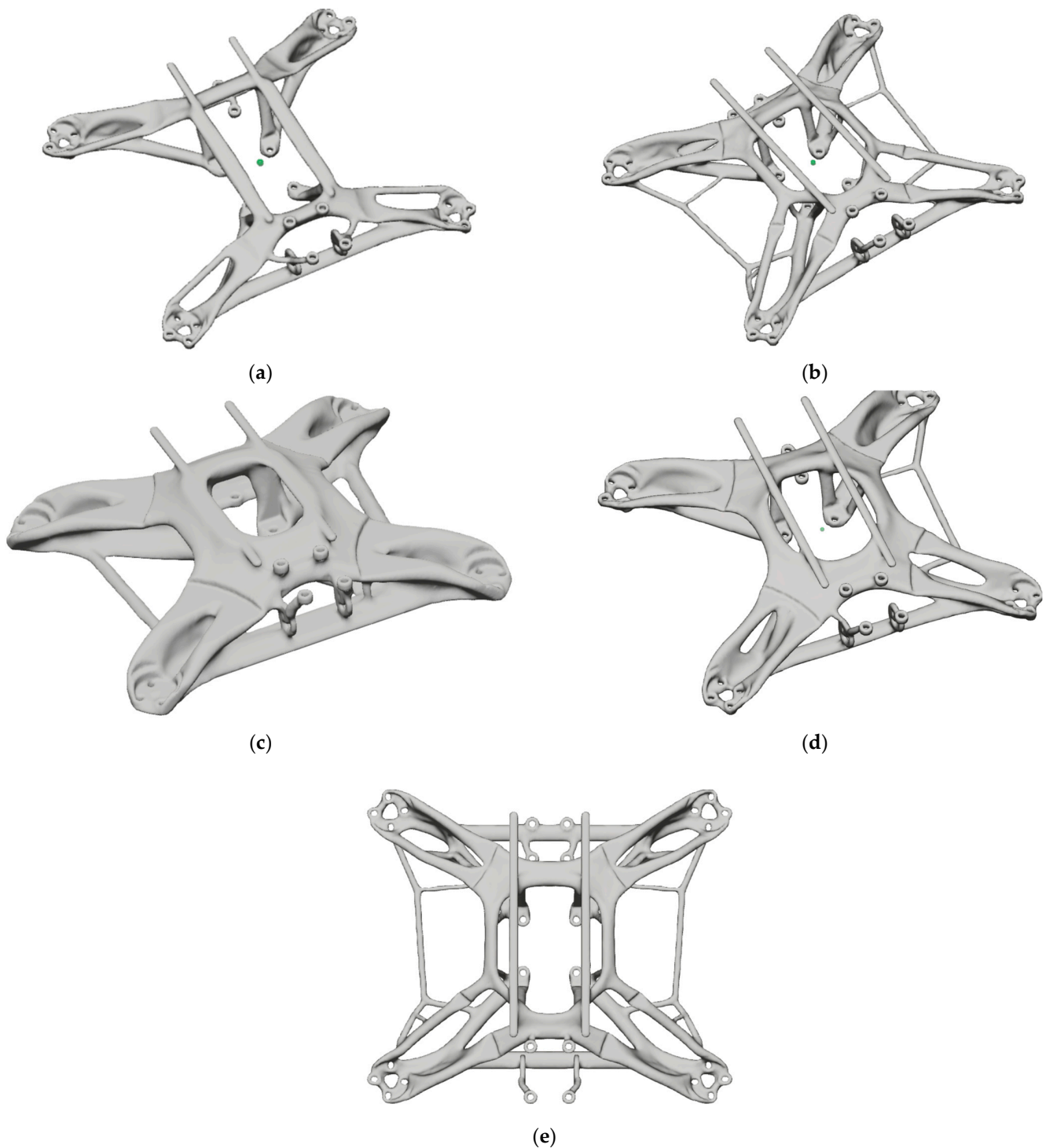


Figure 5. Outcomes achieved through generative design on the quadcopter drone frame: (a) and (b) results achieved by reducing mass; (c–e) results achieved by stiffening the drone frame structure.

4. Simulation Analysis of the Drone Frame

Finite element analysis (FEA) in static analysis on the quadcopter frame is an important step in validating and understanding the structural behaviour of the component prior to manufacturing and physical realization. This finite element analysis was performed to verify that the generative design model will withstand the loads under actual operating conditions.

4.1. Static Analysis of the Drone Frame

For the finite element analysis of the drone frame model, the ANSYS software 2021 R2 (Canonsburg, PA, USA) Static Structural Module was used. In several studies [66–68], 3D-printed materials produced using the FFF process have been considered orthotropic materials. For the FEA simulation in this study, Tough PLA material was considered orthotropic and presented the following properties (Table 7).

Table 7. Orthotropic properties of Tough PLA material used in FEA simulation [64,69,70].

Property	Value
Density [kg/cm ³]	1.22
Young's Modulus X direction [MPa]	2797
Young's Modulus Y direction [MPa]	2797
Young's Modulus Z direction [MPa]	2696
Poisson's Ratio XY	0.33
Poisson's Ratio YZ	0.33
Poisson's Ratio XZ	0.32
Shear Modulus XY [MPa]	1265
Shear Modulus YZ [MPa]	1265
Shear Modulus XZ [MPa]	1235

Based on the mesh convergence analysis, a 1 mm element size was adopted to create a drone frame model comprising 348,368 nodes and 194,446 SOLID 187 tetrahedral elements (Figure 6a). To evaluate the static performance of the drone frame, the following boundary conditions were imposed: the frame was fixed to the surface at the four mounting holes for the FC and ESC and to the battery mounting area (Figure 6b).

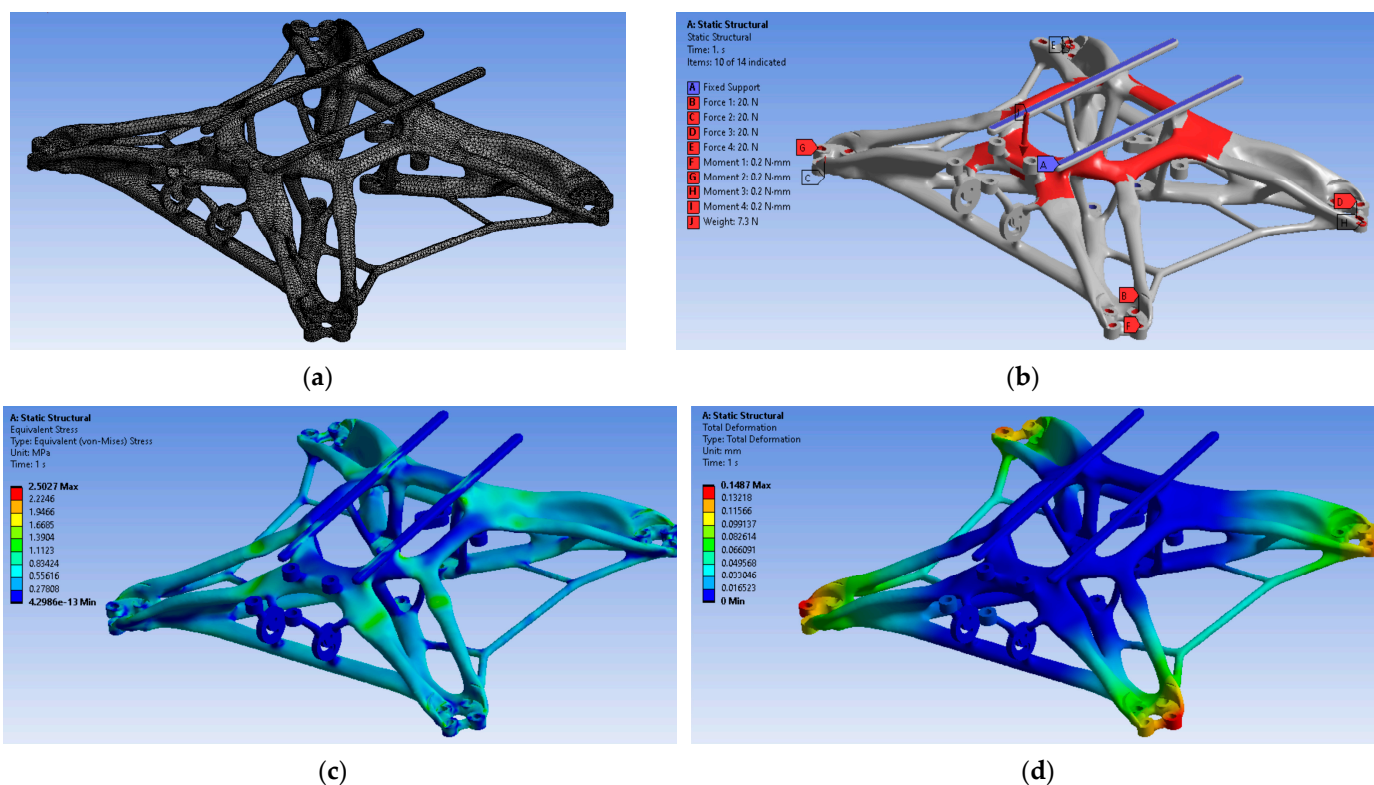


Figure 6. The main stages of finite element analysis of the drone frame: (a) generating the mesh of the drone frame; (b) establishing boundary conditions and loads; (c) distribution of equivalent stresses; (d) distribution of total displacements.

The following loads were applied: a pull force of 20 N was applied to the end of each arm in the area where the motor was mounted; a weight force of 7.3 N was applied to the central surface of the drone frame; and a moment of 0.2 N·m was applied to the surface where the motor was mounted (Figure 6b).

As can be seen in Figure 6c, the equivalent stresses with the highest values are found in the four mounting areas of the FC and ESC, where the model was also fixed. High stresses were also reported in the area between the central section and the battery mounting rails. With an equivalent stress of 2.5 MPa, the frame can withstand the operating stresses of a racing quadcopter drone. However, given that the safety factor is high, at 17, it can be evidenced that the model also could have been 3D printed with an infill density of approximately 50% to reduce weight. The maximum total displacement of 0.14 mm is small (Figure 6d) and occurs at the ends of the arms (in the motor mounting area), but does not affect the structural integrity of the drone.

4.2. Modal Analysis of the Drone Frame

The principal objective of modal analysis is to determine modal parameters, providing a solid basis for evaluating structural system vibrations, diagnosing defects, and optimizing dynamic design [71,72]. For this study, the ANSYS Workbench software system (2021 R2) was used to analyse the quadcopter drone frame. The drone frame was discretized with a mesh size of 1 mm, with 348,368 nodes and 194,446 SOLID187 tetrahedral elements. Considering the operating conditions of the quadcopter drone, an unconstrained free mode analysis was performed.

A modal analysis of an unconstrained drone frame, without any restrictions, is an essential phase in structural engineering and provides fundamental information about the intrinsic behaviour of the analysed structure. As can be seen in Figure 7a,b, the first six vibration modes of the drone frame obtained have a frequency of 0 Hz or very low (in the case of modes 4, 5, and 6) and represent the translational motion on the three axes (X, Y, and Z) and the rotation around the same axes of a rigid body.

Figure 7c–e, and f show the 18 subsequent modes, which describe the vibration characteristics of the drone frame structure. The modal analysis results show that the main areas of vibrational deformation are in the motor mounting zone and at the junction between the arms and the main part of the drone frame.

The first vibration modes correspond to vertical bending loads on the drone's frame arms (mode 7, Figure 7c), with an initial frequency of 95.7 Hz. Subsequently, vibration modes involving a combination of vertical bending and horizontal torsion of the arms, as well as the rods connecting the arms and the accumulator rods, appear (Figure 7d–f). The values of these modes vary between 304.3 Hz and 737.3 Hz. To prevent damage to the drone, it is crucial that the operating frequency of the motors does not coincide with the frame prototype's natural frequency. To verify the operating frequency of the HQProp FLUX 2207 motors used on the drone, the motor speed was determined in rotations per minute (rpm) using Equation (4). Therefore, the speed of the HQProp FLUX 2207 brushless motor is 43,956 RPM.

$$\text{RPM} = \text{KV} \cdot \text{Voltage battery} \quad (4)$$

Equation (5) was used to calculate the operating frequency of the HQProp FLUX 2207 motor. This type of motor has seven magnetic pole pairs (P), and the calculated frequency is 5128 Hz.

$$f_{\text{motor}} = \frac{\text{RPM} \cdot P}{60} \quad (5)$$

The operating frequency of the motor (5128 Hz) is much higher than the maximum natural frequency of the drone frame, as determined by FEA modal analysis (737.3 Hz).

Therefore, resonance should not be a problem for the drone frame obtained through generative design in this scenario. The significant difference in frequency ensures that the vibrations generated by the motors will not amplify to a dangerous degree and cause resonance. In conclusion, the modal analysis of the drone frame demonstrated that there are no natural frequencies within the motors' operating range. Therefore, no problems that could compromise the control or position of the drone are expected to arise from this point of view.

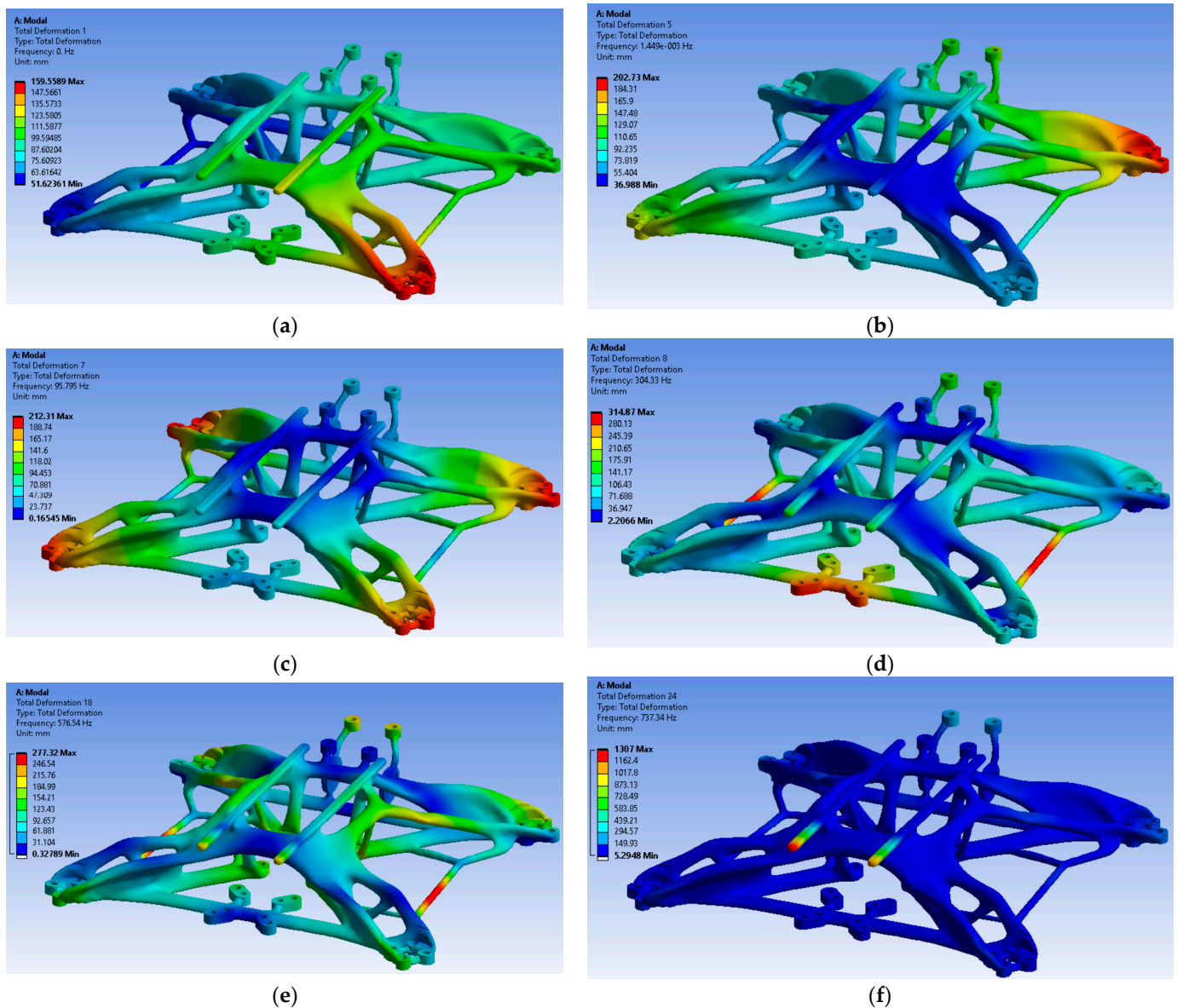


Figure 7. The mode shapes and natural frequencies of the quadcopter drone frame: (a) mode 1; (b) mode 5; (c) mode 7; (d) mode 8; (e) mode 18; (f) mode 24.

5. Additive Manufacturing of Drone Frame

The drone frame, obtained using generative design in the Autodesk Fusion software system, was manufactured using the FFF process with an Ultimaker S5 printer and Tough PLA filament. Preparation for additive manufacturing was carried out using the Ultimaker Cura software system, and the total printing time was 42 h and 25 min, with the mass of Tough PLA filament used being 192 g (Figure 8a). Of course, this filament mass is the total mass, which also includes the support required for the additive manufacturing of the drone frame, and the drone frame has a mass of 135 g (Figure 8b).

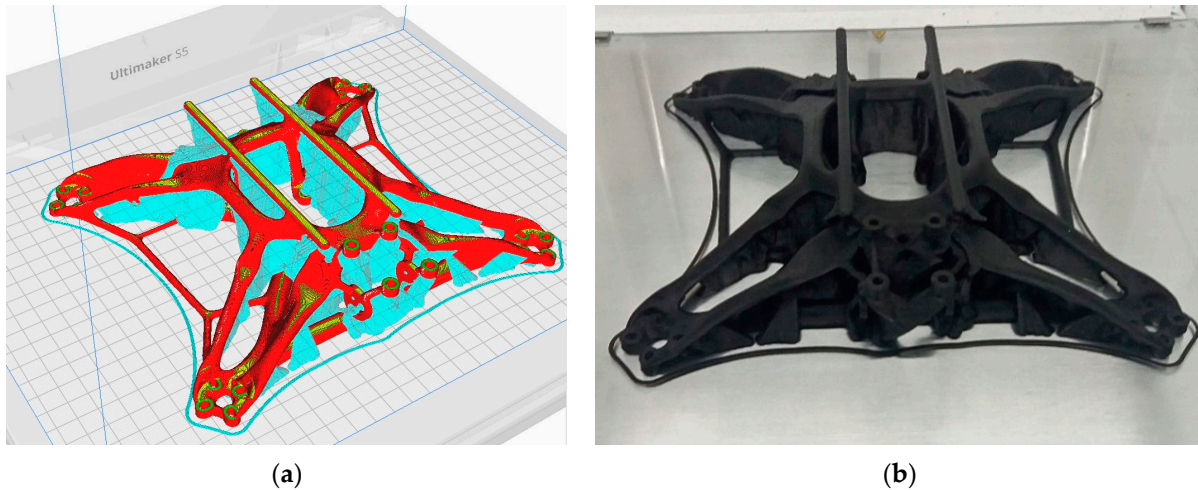


Figure 8. Drone frame manufacturing: (a) preparing the drone frame for manufacturing; (b) 3D-printed quadcopter drone frame.

Setting the 3D printing parameters (Table 8) for the drone frame is an important step because it guarantees a structure that achieves the ideal balance between structural strength, minimal weight, and superior surface quality.

Table 8. FFF process parameters for manufacturing the drone frame.

3D Printing Parameter	Value
Filament	Ultimaker Tough PLA
Filament diameter	2.85
Layer height [mm]	0.1
Infill density [%]	100
Infill pattern	Lines
Build plate temperature [°C]	60
Printing temperature [°C]	215
Printing speed [mm/s]	50
Support structure	Tree
Nozzle diameter [mm]	0.4

6. Assembling Components and Verifying Drone Operation

This section outlines the assembly steps and key component specifications. Figure 9a illustrates all parts, including the remote control and FPV goggles. The first step is to remove the 3D printing supports. Next, mount the FC and ESC in the central area using M3 screws, nuts, and rubber gaskets to minimize vibration (see Figure 9b).

The FPV camera is mounted using two M2 screws to transmit images to the goggles (Figure 9c). Next, the video transmitter and antenna are secured with four M3 screws at the rear of the drone (Figure 9d). The radio receiver and its antenna are mounted in the central area at the front, under the GoPro camera mount. Next, mount the four motors, each with its own propeller (Figure 9e). Finally, the battery is attached with two Velcro strips to the frame rails (Figure 9f).

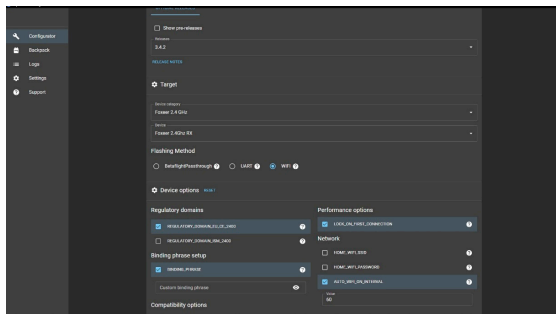
Next, the drone components are initialized to prepare it for flight. First, once the drone has been assembled and all the components connected, it must be connected to a computer via a USB cable. Most of the settings for the quadcopter drone are configured using the BetaFlight Configurator software, version 10.10.0.



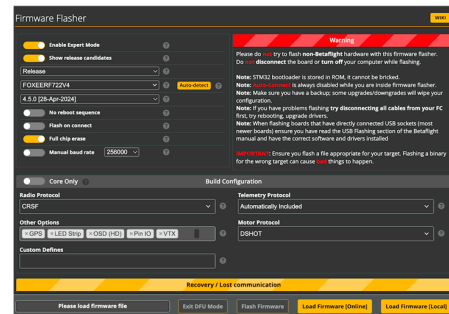
Figure 9. Stages of the quadcopter drone assembly process: (a) drone components; (b) mounting the FC and ESC in the central area of the frame; (c) mounting the FPV camera; (d) mounting the video transmitter and associated antenna; (e) mounting the motors and propellers; (f) mounting the battery.

The verification process for the drone started with the installation of firmware on the radio receiver and flight controller (FC) via the ExpressLRS Configurator application (Figure 10a,b). Next, the accelerometer was calibrated (Figure 10c), the UART ports were set up for communication, and the gyroscope and accelerometer frequencies were adjusted,

along with the signal loss alarm (Figure 10d). Battery operating limits were set to prevent wear and tear, and an OSD warning was added (Figure 10e). Radio control commands were then tested, followed by configuration of the motors' rotation direction (Figure 10f). The OSD profiles were also customized to display essential information during flight (Figure 10g). Finally, the firmware was installed on the radio control (Figure 10h), after which functional and low-altitude flight tests were performed.



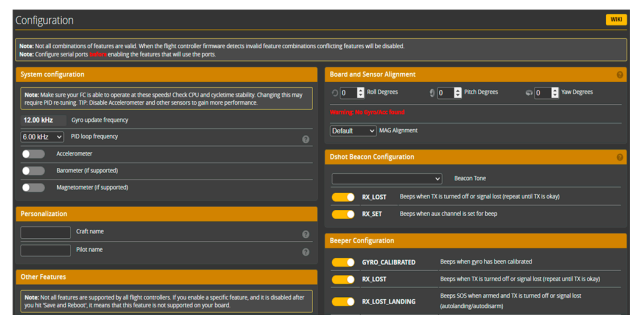
(a)



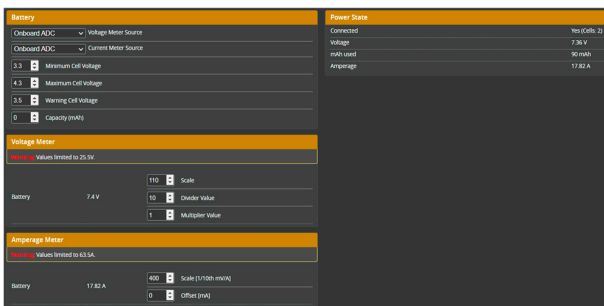
(b)



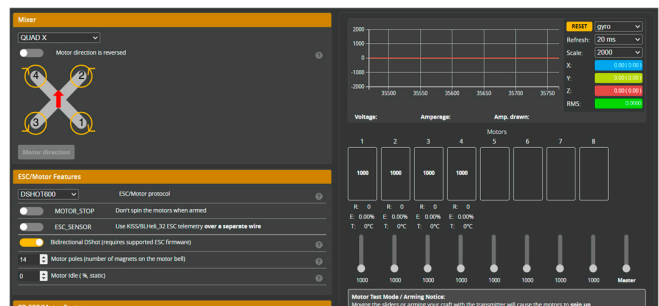
(c)



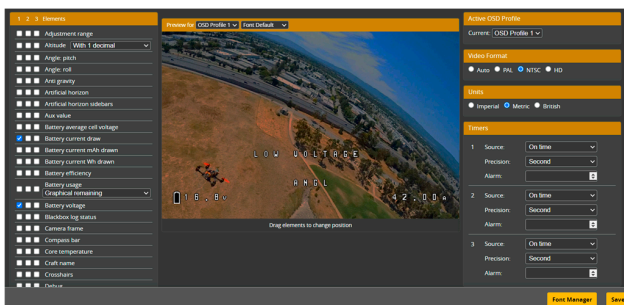
(d)



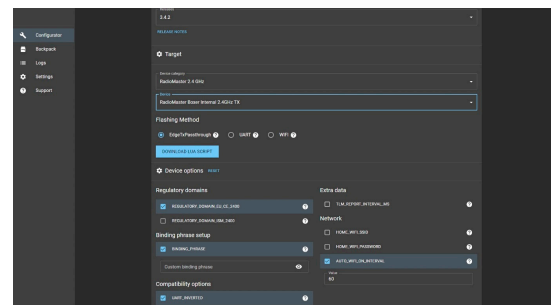
(e)



(f)



(g)



(h)

Figure 10. Steps in the quadcopter drone operation checking process: (a) ExpressLRS Configurator–radio receiver firmware update; (b) FC firmware update; (c) accelerometer calibration; (d) gyroscope and accelerometer frequency setting; (e) setting battery alerts; (f) setting the motor rotation mode; (g) setting the OSD profiles; (h) radio control firmware update.

7. Conclusions

This study demonstrates progress in the additive technology and optimization of aerospace structures for UAVs. However, there is a gap in the integration of additive manufacturing technology and generative design methods in the manufacture of UAV structures. This study proposes to reduce this gap by combining these approaches to create higher-performance UAV structures. Given the importance of developing optimal structural designs for UAVs, this study is relevant and has the potential to contribute significantly to improving efficiency, performance, and innovation in the UAV industry and related industries in the future. The following conclusions can be drawn from this study:

- The advantages of generative design combined with additive processes are undeniable: reduced weight (18% reduction compared to the original carbon fibre version), design freedom, reduced material consumption, short manufacturing times and low costs, and a wide range of materials. This reduction in mass can be attributed to the low density of the Tough PLA material (1.22 g/cm^3) and from the additively manufactured frame of the quadcopter drone. However, at present, the additive manufacturing technology required for a complex geometric structure such as a quadcopter drone frame is not yet sufficiently developed to be reliable for medium- or large-scale production, making it ideal for small-scale production and prototyping.
- Adaptability through the ability to create structures for different specifications (load values, structure shapes, or safety factors) that would otherwise be impossible to manufacture is an advantage over components made using conventional manufacturing methods. Another advantage is that the resulting quadcopter frame is a single component and does not require assembly like commercially available ones.
- The drone frame was quickly and easily prototyped using additive manufacturing processes involving extrusion of the material. Unlike traditional methods, such as CNC machining or injection moulding, FFF processes have several limitations in terms of production scalability. These include reduced cost and time efficiency at high volumes, low production speed, and inferior mechanical strength compared to composite materials. Conventional methods could not be used in this case due to the complexity of the model. However, although the material used is not as strong as a composite or metal material, it is strong enough to meet the criteria required for the design and operation of the quadcopter drone.
- Frame maintenance is difficult because Tough PLA filament requires a dry and cool environment to prevent rapid deterioration, which is facilitated by high humidity, high temperature, and prolonged exposure to UV rays. Usually, damage to the frame through cracking or breaking requires replacement with a new one, as repair is not possible or feasible.
- Finite element analysis (FEA) and functional testing have confirmed that the frame structure is sufficiently strong to withstand static and dynamic loads. Combined with significantly reduced production costs, these results validate the performance of the frame, which is representative of similar drones on the market.

In conclusion, this paper has demonstrated the feasibility of building a drone by completing all the stages required for the development of a UAV aeronautical product (preliminary design–generative design–finite element analysis–3D printing–assembly and operational testing). Future research will include scaling up the drone, changing the drone material (continuous fibre filament or high-strength polymers), adding a new mission with the ability to transport various products (medicines, emergency equipment, or even food), or using it for various inspections by adding sensors (gas detection and air monitoring), and increasing operational capability through the use of more powerful batteries and a ground control station.

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