
PID BASED HIERARCHICAL AUTONOMOUS SYSTEM PERFORMANCE MAXIMIZATION OF A HYBRID UNMANNED AERIAL VEHICLE (HUAV)

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ABSTRACT

In this article, preliminary and conceptual design, dynamic modeling and flight dynamics modes of an unmanned aerial vehicle which have both helicopter flight mode and fixed-wing aircraft flight mode are investigated. The air vehicle (ZANKA-IV) which is planning to be produced in Erciyes University Faculty of Aeronautics and Astronautics Drone Laboratory captures both passive and active morphing properties. Furthermore, this hybrid unmanned aerial vehicle (HUAV) will be controlled by proportional–integral–derivative (PID) controller based autopilot system. In this article, sizes of morphing apparatuses and autonomous system are determined simultaneously in order to maximize autonomous flight performance for the first time. Also, a stochastic optimization method is applied for this simultaneous design. As a result of that, the optimum solution for design parameters is obtained very fast and accurately. Thus, as a result of this approach, autonomous control of HUAV becomes noticeably efficient. Finally, 10% improvement on autonomous performance index is achieved, and autopilot successfully tracks a reference trajectory with pitching angle of 30 degrees.

Keywords: Hybrid UAVs, Autonomous flight, Autonomous performance maximization

1. INTRODUCTION

With technological advances in recent years, unmanned aerial vehicles (UAVs) has become increasingly popular in the field of academic research and engineering applications [1-4]. UAVs are performed in many situations of surveillance and analysis such as military applications, airborne mine detection, digital mapping, electronic warfare, radar, saturation, and mixing, recovery, meteorology, aerial photo and video capture, traffic surveillance, small package transportation, scientific research, illegal construction as a fast and secure method [1-4]. These vehicles can be supplied by electric power or liquid fueled, and their height may be as small as a bug while their weight may be large enough to handle thousands of kilograms according to the field of application. As a result of their wide application area, there are various criteria for classification of UAV systems according to flight range, altitude and hover period. UAVs are separated as HALE (high altitude, long endurance UAVS), MALE (medium altitude, long endurance), EN-TUAV (Tactical UAV for long duration cruise, TIHA), LR-TUAV (long range tactical UAV), MR-TUAV (mid-range), SR-TUAV (short-range), CR-TUAV (close range), MUAV (mini), MAV (micro UAV), and vertical take-off and landing (VTOL) UAV [5]. In this article, electric powered and both active and passive morphing CR-VTOL UAV is modeled and simulated. Similarly, IAI Heron [6], served in the Israel army, is known as the most advanced HUAV in the world. However, this HUAV has not any morphing capability. For this reason, it is not possible to run with optimum performance. A small number of HUAV studies also concluded in Turkey (see [7]).

During hover, efficiency of VTOL aircraft is less than the one of helicopter due to the smaller efficient rotor area. However, efficiency of VTOL aircraft during cruise flight is much better than the one of

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helicopters. As an aircraft spend much more time in cruise flight than hover, overall performance of VTOL aircraft is better than helicopter's performance (see [7-8]). Also, VTOL aircrafts are capable of carrying much more load than same size helicopters. Additionally, performance of a helicopter is limited by energy conversion efficiency and lack of resistance to high cruising speed at rotor tips. In contrast to helicopter, VTOL aircraft is not suffered from cruise flight limitations. VTOL UAVs have the ability of both vertical take-off and landing and flying high cruising speeds. Thus, VTOL UAVs are able to take-off and land in hazardous environments even. Additionally, VTOL UAVs can reach the target work area in a very short time. All these features allow VTOL UAVs to operate with high performance in a wider range of tasks compared to standard and conventional unmanned helicopters as seen in Figure 1.

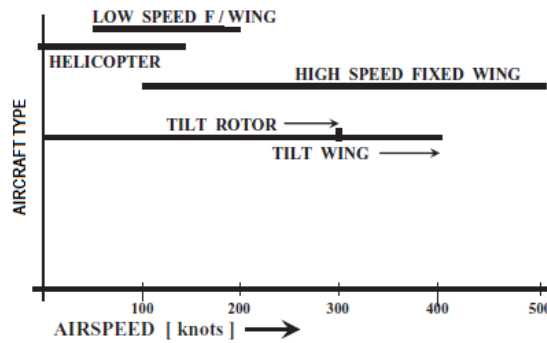


Figure 1. Speed ranges of different aircraft types

For designing an UAV control system, autopilots that help to control UAVs automatically without human assistance during the flight are very important. Autopilots are integrated intelligent systems, and consist of state sensors and controllers. State sensors measure different parameters of UAV continuously by using several sensors such as Global Positioning System (GPS), accelerometer, magnetometer, gyro, and pitot-tube. Control systems use these measurements and calculate differences between current and required conditions.

In this article, a VTOL HUAUV called ZANKA-IV that has both passive and active morphing features is modeled at Faculty of Aeronautics and Astronautics, Erciyes University. The drawing of our HUAUV is given in Figure 2. Also, dynamic modes of the aircraft are compared to the literature; applied autopilot is mentioned briefly, and the sizes of the morphing mechanisms and PID parameters are optimized by a stochastic optimization method. Moreover, closed-loop simulation results, and improvements on autonomous performance are proposed. A summary of this study was presented and published in a conference [9].

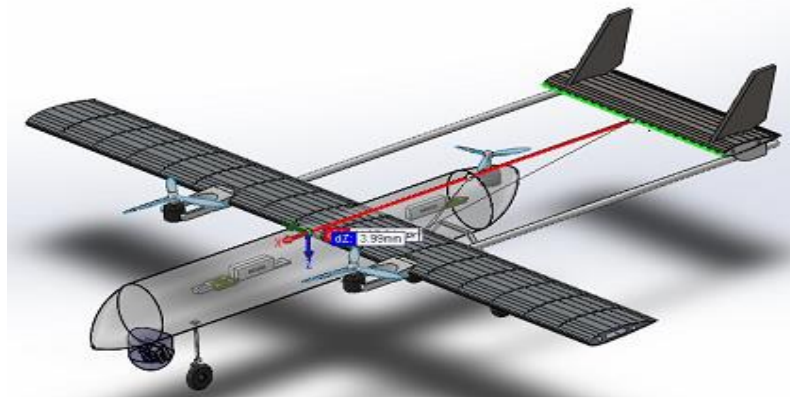


Figure 2. Drawing of the HUAUV (ZANKA-IV)

In this paper, a HUAV is equipped with passive and active morphing features, for the first time in the literature. Also, for the first time, sizes of passive and active morphing mechanisms which let the aircraft better autonomous performance is designed to help maximize autonomous performance. Passive and active morphing provide the some changes in HUAV parameters (e.g., wing, tailplane, etc.). The significant difference between active and passive morphing is that active morphing requires continuous measurements and actuation while passive morphing changes are before the HUAV begins the flight mission or at a separate time during flight. Passive and active morphing parameters as well as the automatic flight control system are optimized simultaneously. For this purpose, simultaneous perturbation stochastic optimization (SPSA) which is an adaptive stochastic optimization method is applied first time. In our HUAV the passively morphing parameters are position of wing and tailplane in longitudinal axis and the actively morphing parameters are extension ratios of wing and horizontal tail (see [10] for details).

2. MATERIALS AND METHODS

2.1. HUAV Dynamics and State-Space Model

After detailed dynamic modeling process, equations (1) and (2) are presented as longitudinal and lateral state-space parametric models of HUAV, respectively. These equations are also suitable for various HUAV geometries. Reference [11] can be referred to decide the value of parameters in the matrix. The detailed computation of each parameter is proposed in this reference. These state-space models can be examined for the all flight modes movements except extreme nonlinear behaviors of HUAV. Although there are various references for aircraft modes in the literature such as [11], there is no clear presentation for helicopter state-space modes. In equation (1): u , w , q and θ indicate longitudinal, vertical and angular velocity of pitching motion, and pitch angle of UAV, and δ_T and δ_e are throttle and elevator control, respectively. In equation (2): v , p , r and ϕ denote lateral and angular velocity of rolling motion, angular velocity of yawing motion, roll angle of UAV, and δ_a and δ_r are aileron control and rudder control, respectively.

$$\dot{\mathbf{x}}_l = \underbrace{\begin{bmatrix} X_u & X_w & 0 & -g & 0 \\ Z_u & Z_w & u_0 & 0 & 0 \\ M_u + M_w Z_w & M_w + M_w Z_w & M_q + M_w u_0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ -\sin(\theta_0) & \cos(\theta_0) & 0 & -u_0 \cos(\theta_0) & 0 \end{bmatrix}}_{A_l} \mathbf{x}_l + \underbrace{\begin{bmatrix} X_{\delta_T} & X_{\delta_e} \\ Z_{\delta_T} & Z_{\delta_e} \\ M_{\delta_T} + M_w Z_{\delta_T} & M_{\delta_e} + M_w Z_{\delta_e} \\ 0 & 0 \\ 0 & 0 \end{bmatrix}}_{B_l} \mathbf{u}_l \quad (1)$$

$$\dot{\mathbf{x}}_{la} = \underbrace{\begin{bmatrix} \frac{Y_\beta}{u_0} & \frac{Y_p}{u_0} & -(1 - \frac{Y_r}{u_0}) & -\frac{g}{u_0} \cos(\theta_0) & 0 \\ L_\beta + \frac{I_{xz}}{I_{xx}} N_\beta^* & L_p + \frac{I_{xz}}{I_{xx}} N_p^* & L_r + \frac{I_{xz}}{I_{xx}} N_r^* & 0 & 0 \\ N_v^* + \frac{I_{xz}}{I_{zz}} L_v^* & N_p^* + \frac{I_{xz}}{I_{zz}} L_p^* & N_r^* + \frac{I_{xz}}{I_{zz}} L_r^* & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & \sec(\theta_0) & 0 & 0 \end{bmatrix}}_{A_{la}} \mathbf{x}_{la} + \underbrace{\begin{bmatrix} 0 & \frac{Y_{\delta_r}}{u_0} \\ L_{\delta_a}^* + \frac{I_{xz}}{I_{xx}} N_{\delta_a}^* & L_{\delta_r}^* + \frac{I_{xz}}{I_x} N_{\delta_r}^* \\ N_{\delta_a}^* + \frac{I_{xz}}{I_{zz}} L_{\delta_a}^* & N_{\delta_r}^* + \frac{I_{xz}}{I_{zz}} L_{\delta_r}^* \\ 0 & 0 \\ 0 & 0 \end{bmatrix}}_{B_{la}} \mathbf{u}_{la} \quad (2)$$

Longitudinal and lateral flight dynamic modes of modeled ZANKA-IV that will be manufactured at Faculty of Aeronautics and Astronautics are shown in Figure 3. These modes are qualitatively and quantitatively similar to the UAV modes in the literature (see [12-14]). Thus, our modeling process is validated.

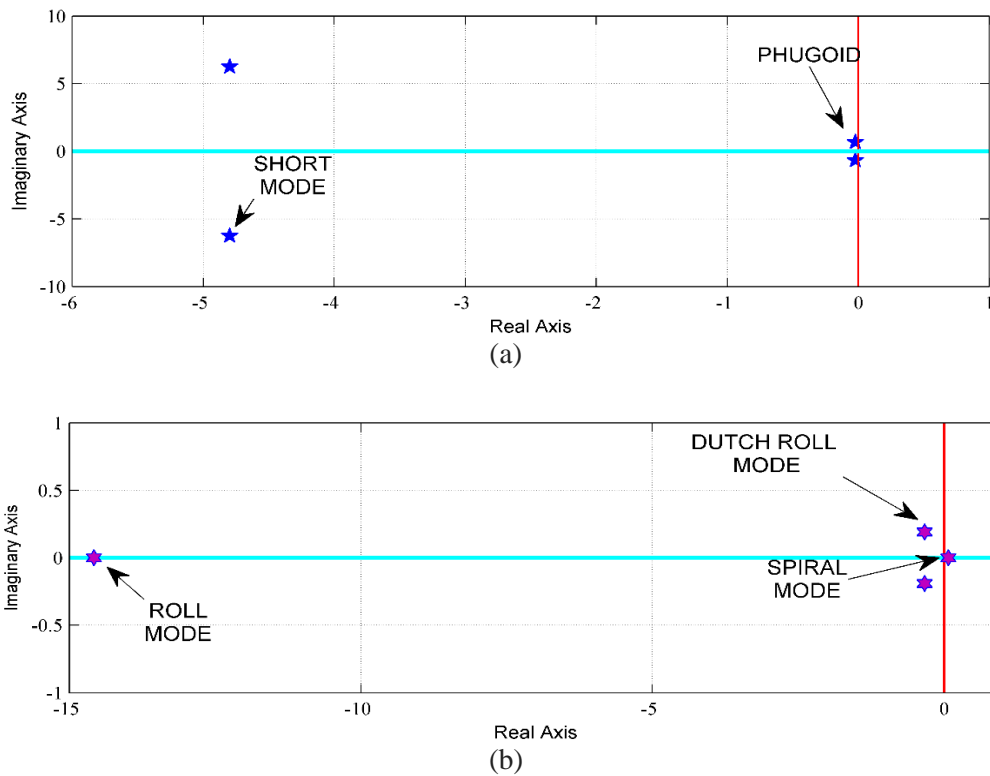


Figure 3. (a) Longitudinal and (b) lateral flight dynamic modes of ZANKA-IV

2.2. Hierarchical PID Based Autopilot Structure

In most general form, PID-based hierarchical autopilot system provides height, yaw angle and velocity tracking. Autopilot system includes six PID controllers in its three layers. These controllers provide trajectory tracking. If users need to utilize all of them, 18 parameters are required to be tuned. However, in this simultaneous design study, there are only two additional structural parameters (assembly points of wing and tail to fuselage). Therefore, the block diagram of the commercial autopilot used in this study is illustrated in Figure 4. More detailed information can be found in References [15, 16].

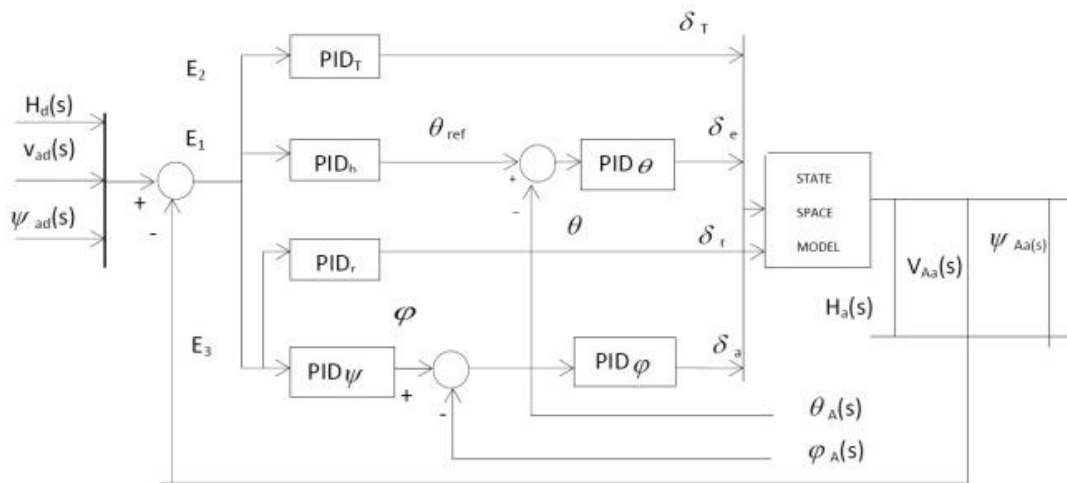


Figure 4. Detailed Block Diagram of the Hierarchical Commercial Autopilot

It is seen from this block diagram that the hierarchical autopilot system has 6 PID controllers. This autopilot is available to track altitude, velocity, and yaw angle. The T , h , r , ψ , θ , and ϕ refer throttle, altitude, rudder, yaw, pitch and roll for PID indices.

2.3. Problem Formulation

PID-based hierarchical autopilot system track yaw angle, altitude and velocity generally. Commercial autopilot system accommodates 6 PID controllers in 3 layers (outer, middle and inner layers). These PID controllers have upper and lower limits on the P (proportional), I (integral) and D (differential) parameters. Appropriate estimation of these parameters provides a favorable trajectory tracking. If the autopilot is requested to track yaw angle, altitude and velocity, 18 controlling parameters should be chosen appropriately. In this study, the PID parameters are estimated by SPSA to minimize a cost function. The cost function is consisting of some parameters related to the trajectory tracking (i.e. settling time, rise time and overshoot). If this function becomes minimum, the tracking errors become minimum. Performance means this reality. It is hereby accepted that if performance becomes maximum then energy consumption also reduces. Energy saving is considered with respect to the default initial conditions that $P=50$, $I=5$, $D=50$ for PID controller, %50 spanwise extension for both wing and horizontal tail, and initial longitudinal positions of wing and horizontal tail.

2.4. Stochastic Optimization Method (SPSA)

In this paper, SPSA is applied for simultaneous design. According to other optimization methods, SPSA has many advantages. Especially, for highly complicated optimization problems in which the gradient and Hessian is not possible to be calculated analytically (such as in this study), SPSA is able to estimate the optimal solution. With this aspect, it addresses shortcomings of deterministic optimization methods (e.g., golden section search method). It also has the advantage over other stochastic optimization methods. For instance, while genetic algorithms required to calculate 2^n (n is the optimization variables) to compute gradient for each iteration, SPSA calculate the gradient just in two steps. Simply, when the number of optimization variables is increased such as to 4, genetic algorithm makes 16 calculations for optimization at each iteration while SPSA makes 2 calculations only. For this reason, SPSA is much faster to converge an accurate solution. On the other hand, when there are even constrains on the optimization parameters, SPSA can successfully reach optimum solutions.

Because of the very complicated relation between the constraints on the optimization variables and the cost function (J , see equation 3), the derivatives of the cost function with respect to these variables is not possible to be computed analytically. For this case, stochastic optimization methods are needed to solve this problem. In this study, SPSA is selected because of its previously performance on solving similar constrained optimization problems (see [17-19]).

$$J = \sum (T_{st} - T_{st_u})^2 + (T_{rt} - T_{rt_u})^2 + (\%OS - \%OS_u)^2 \quad (3)$$

where T_{st} indicates settling time, T_{rt} is rise time, %OS demonstrates overshoot percentage, and u means upper value of these parameters. Terms of cost function is calculated as below:

$$\begin{aligned} \text{If } T_{st} \geq T_{st_u}, \quad T_{st} \text{ is non-defined else} \\ T_{st} < T_{st_u}, \quad T_{st} \text{ is its value} \end{aligned} \quad (4)$$

$$\begin{aligned} \text{If } T_{rt} \geq T_{rt_u}, \quad T_{rt} \text{ is non-defined else,} \\ T_{rt} < T_{rt_u}, \quad T_{rt} \text{ is its value} \end{aligned} \quad (5)$$

If $\%OS \geq \%OS_u$, $\%OS$ is non-defined else

$$\%OS < \%OS_u, \quad \%OS \text{ is its value.} \tag{6}$$

2.5. Design Results


Some geometrical data of ZANKA-IV are shown in Table 1. According to the studies on existing UAVs and consideration of user requests, ZANKA-IV is designed to be weight of 12 kg, carrying a payload weight of 1.5 kg, and endurance of 1.5 hours.

Table1. Initial HUAV Data (ZANKA-IV)

Wing	Span	Chord Length	Max. Thickness	Mass	Materials used
	$b_w = 250$ cm	$c_w = 25$ cm	$\delta_w = 39.27$ mm	3487.0 gr	Styrofoam, Balsa, Carbon Composite, Carbon Fiber
	Aspect Ratio	Taper Ratio	Dihedral	Swept Angle	Airfoil
	10	1	0 degree	0 degree	NACA 2415
Horizontal Tail	Span	Chord Length	Max. Thickness	Mass	Materials used
	$b_{HT} = 80$ cm	$c_{HT} = 20$ cm	$\delta_w = 33.82$ mm	693.63 gr	Styrofoam, Balsa, Carbon, Composite, Carbon Fiber
Vertical Tail	Height	Length (rudderless)	Length (including rudder)	Mass	Materials used
	25.38 cm	24.96 cm	31.55 cm	467.37 gr	Styrofoam, Carbon, Composite, Carbon Fiber
Fuselage (Engine, landing gear, ESC, autopilot, etc. included)	Length	Max. Depth	Max. Width	Mass	Materials used
	155.32 cm	30.0 cm	30.0 cm	5852.0 gr	Styrofoam, Balsa, Carbon Composite, Carbon Fiber, Plywood

The similar HUAVs in the world are compared to ZANKA-IV in Table 2.

Table 2. Comparison of HUAVs

HUAV	 Panther [6]	 Mini-Panther [6]	 TURAC [7]	 ZANKA-IV (Erciyes)
Fuel Type	Electric	Electric	Electric	Electric
Endurance	6 h	1.5 h	1.42 h	1.5 h
Payload	8.5 kg	1.5 kg	8.0 kg	1.5 kg
Total weight	65.0 kg	12.0 kg	47.0 kg	12.0 kg
Mass of Morphing Mechanism	X	X	X	0.5 kg
Passive Morphing	X	X	X	✓
Active Morphing	X	X	X	✓

After estimating PID parameters by SPSA, tracking trajectory of simultaneously designed ZANKA-IV for pitching angle of 30 is given in Figure 5. The exact and relatively improved autonomous performances through simultaneously design are shown in Figures 6 and 7.

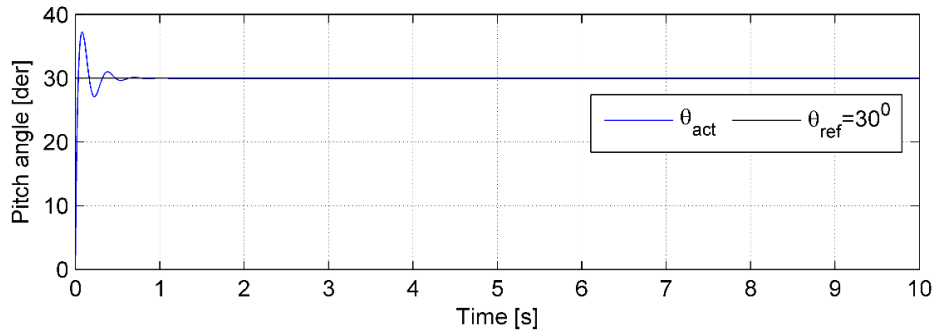


Figure 5. Tracking trajectory for pitching angle through the final optimum solution in the condition without turbulence

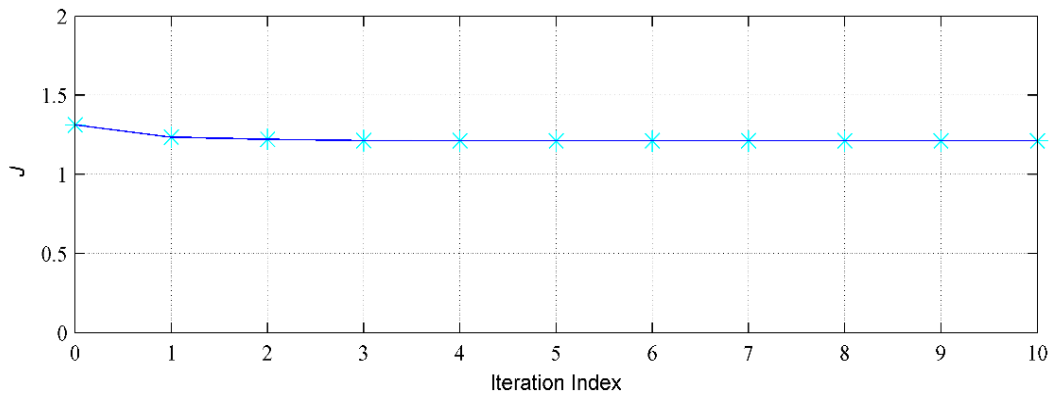


Figure 6. The energy/performance behavior in the optimization steps

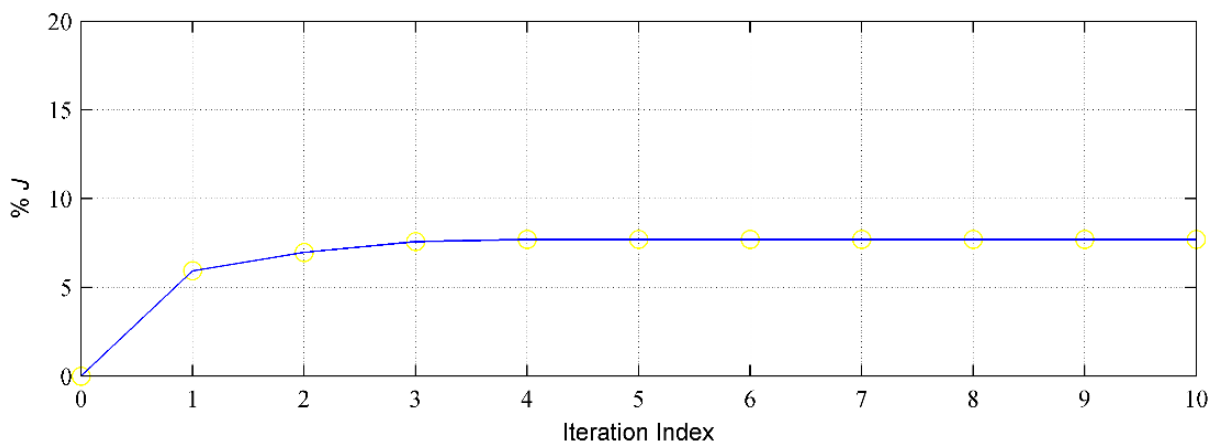


Figure 7. The relative energy/performance improvement behavior in optimization steps

In Figure 8 a brief illustration of morphing mechanism is given. Its working principle can be summarized as follows: Our morphing wing or horizontal tail includes reductive micro motor, two equal number of gear teeth, M3 worm gear and bidirectional brushing ESC. A micro motor is a DC motor operating between range of 6 V to 9 V. A gear motor is mounted to the shaft of this motor. This gear runs

together with a second gear mounted on a bearing. Interior of the gear mounted on bearing has M3 tooth which is in opened state. Signal sent to ESC through a bistable switch defined on the remote control gives the engine right or left turn. Motor, and therefore mechanical system which move with respect to the direction of rotation of gears lead worm screw in-or-outward. This movement results by moving wing tips parts, produced by outer mold method, over main section. As a result, it provides the change of the wing or horizontal tail area (see [20] for more details).



Figure 8. Illustration of morphing mechanism

3. CONCLUSION

In this article, initially, preliminary design, dynamic modeling and flight modes of an unmanned air vehicle called ZANKA-IV that has both fixed-wing and helicopter flight mode are investigated. ZANKA-IV which will be manufactured at the Drone Laboratory settled at Erciyes University has passive and active morphing mechanisms. Our HUAV will be controlled automatically by PID-based autopilot system. In this study, the size of morphing mechanisms and autonomous system parameters are firstly simultaneously considered to maximize the autonomous performance of a Hybrid UAV. For this simultaneous design, a stochastic optimization method is utilized to optimize the design parameters. This algorithm provided to converge to the optimum solutions fast. Moreover, through this design approach, a significant amount (approximately 10%) of energy saving is achieved with respect to the default initial conditions (i.e. $P=50$, $I=5$, $D=50$ for PID controller, %50 spanwise extension for both wing and horizontal tail, and initial longitudinal positions of wing and horizontal tail) by autonomous control of the HUAV. Consequently, the pitching angle of 30 degrees is successfully tracked in simulation environment.

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