



Design and FPGA Implementation of Immune-PID Controller Based on BBO Algorithm for Heart Rate Regulation

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Abstract: In this paper an optimal Immune-Proportional-Integral-Derivative (Immune-PID) scheme is suggested to control to the heart rate of humans and regulate it, based on Yanagihara, Noma, and Irisawa (YNI) as mathematical model of the heart rate. The parameters of the suggested immune PID controller are optimized by using Biogeography-Based-Optimization (BBO) algorithm. In addition, the proposed controller scheme is implemented using Field Programmable Gate Array (FPGA) due to its great data storage capacity, low consumption of energy, and high speed of operation. The Xilinx system generator blocks and Spartan-6 (XC6SLX45T-3FGG484) board are used due to availability with 1% utilization of hardware platform slices. The performance of the proposed controller is compared to traditional PID and immune controllers. That comparative analysis results show best improvement when using the proposed controller with 0% maximum overshoot, a reduction of steady state error and rising time reaches to 0.1 and 0.0023 second respectively and faster response reaches to 0.01 second.

Keywords: Use PID controller, Immune system, Biogeography-based optimization, YNI model, FPGA.

1. Introduction

Disease of the heart and blood vessels make up the majority of death causes around the world, mostly in advanced countries. There are two famous heart-related conditions, slow heartbeat (bradycardia) and too-fast heartbeat (tachycardia), leading to fainting or shortness of breath and sometimes causing death [1]. When the heartbeat drops less than 60 bpm it classified as Bradycardia, its signs of sickness become clear when the HR drops below than 50 bpm, and when the HR exceeds 120 bpm is classified as tachycardia [2]. In order to treat these two cases, there are several solutions, including simple therapeutic drugs, or implant a pacemaker which is an electronic system that monitors cardiac functions, and compensate for any disorder.

Researchers have developed pacemaker control systems, especially in recent years, Aabid et al. use a traditional PID controller based on Ant-Colony Optimization algorithm to verify PID coefficients to regulate the HR [3]. Arunachalam et al. suggest a

robust fractional-order PID controller for the cardiac pacemaker, which is tuned by the Ziegler-Nichols method [4]. Govind and Sekhar present a PID control system for HR regulation as pacemaker in an efficient way. It is executed via a loop consisting of an adaptive correction factor, the MIT rule, and Delta rule. The Simulation results claim that the suggested PID controller provides better performance than traditional PID controller [5]. Castroa and Malathi suggest a classical PID controller with back-propagation neural network, for heart rate regulation [6]. the paper uses Zeigler-Nichols tuning to determine the PID parameters, where it represents the output layer weights in the neural network. M. E. Karar suggests a new back-stepping adaptive controller for heart rate regulation, in dual sensor pacemaker based on radial basis function [7]. The developed controller is tested using pre-determined data of four patients heart rate during different activities. A Mamdani fuzzy proportional derivative controller is proposed in [8], to improve the cardiac pacemaker and regulate the heart rate. The controller

represents an integration of the traditional PID. To obtain the optimal gains of the controller an optimization technique is used. The authors previously presented three structures for immune PID controller [9], to regulate the HR. The differential evolution algorithm is adopted to optimize the controller parameters, the best performance is obtained from the third structure. A modified model-free sliding mode controller is presented in [10] for HR regulation in a pacemaker, the BBO algorithm is used to tune the parameters of the controller.

Although the mentioned controllers are claimed to have good performance, but these controllers use only PID approaches to improve the performance of the pacemaker, with few controllers used as pacemaker. There is no attempt to design or implement these controllers, FPGA to combine it as a hardware with the pacemaker. According to above reasons, the main aim of this paper is to design and implement an intelligent controller as a pacemaker for heart rate regulation, where the Immune-PID controller is used to regulate the abnormalities, which occurs in the HR, the BBO algorithm is used to improve the controller parameters. To accomplish these objectives, YNI mathematical model is considered here, and implemented using FPGA (Spartan-605), due to its great datastorage capacity, low energy consumption, and high speed of operation. The rest of this paper is structured as follows: The YNI mathematical model of HR is presented in Section 2. The details of the Immune system are described in Section 3, while the design of the Immune-PID controller is explained in Section 4. The BBO algorithm is illustrated in Section 5. The FPGA implementation is addressed in Section 6. The proposed controller's analysis and simulation results are discussed in Section 7, while the final conclusions are listed in Section 8.

2. YNI Heartbeat model

The YNI mathematical model represents the action potential behavior of the Sino-Atrial cells and considered from its earliest models [1, 11]. The YNI model type is Hodgkin-Huxley, which describes standings of conductance and ionic-currents. It can simulate the action potentials of heart accurately, which based on the data of voltage-clamp, taking into consideration the relationship between current and voltage, and simulates the unprompted action potential. It contains one time-independent current called the leakage current I_l , and four time-dependent currents (potassium current I_k , hyper-polarization delayed inward current I_h , sodium current I_{Na} , and slow inward current I_s) as in Eq. (1). Fig. 1 shows an

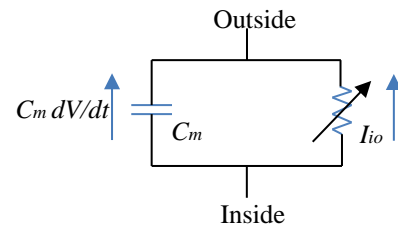


Figure. 1 The equivalent electrical circuit of cell membranes model [11]

electrical circuit equivalent to the cell membrane model. The $\mu A/cm^2$ is the transmembrane current unit [11, 12].

$$C_m \frac{dV}{dt} + I_l + I_k + I_h + I_{Na} + I_s = I_{app} \quad (1)$$

where C_m ($\mu F/cm^2$) refers to the cell membrane capacitance and it is a constant; I_{app} is the externals applied current. Eq. (1) can be rewritten to be:

$$C_m \frac{dV}{dt} + I_{ion} = C_m \frac{dV}{dt} + C_m \frac{V}{R_m} = I_{app} \quad (2)$$

where R_m and C_m denote the cell membrane resistance and capacitance respectively, where:

$$I_l = 0.8 \left(1 - \exp^{-\frac{V+60}{20}} \right) \quad (3)$$

$$I_k = 0.7p \frac{\exp(0.0277(V+90))-1}{\exp(0.0277(V+40))} \quad (4)$$

$$I_h = 0.4q(V + 45) \quad (5)$$

$$I_{Na} = 0.5m^3h(V - 30) \quad (6)$$

$$I_s = 12.5 (0.05 + 0.95d)(0.05 + 0.95f) \left(\exp^{\frac{V-10}{15}} - 1 \right) \quad (7)$$

$$\frac{dw}{dt} = \alpha_w(1 - w) - \beta_w w \quad (8)$$

and w is $\{d, q, h, p, m, \text{ or } f\}$.

The gating variables $d, q, h, p, m,$ and f must fulfill Eq. (8) and Eq. (9) with constant values listed in Table 1, which provides the values of the constants α_w and β_w .

$$\alpha_w, \beta_w = \frac{C_1 \exp\left(\frac{V-V_0}{c_2}\right) + C_3(V - V_0)}{1 + C_4 \exp\left(\frac{V-V_0}{c_5}\right)} \quad (9)$$

The parameters ($\alpha_q, \beta_q, \alpha_d, \alpha_p, \beta_f$) which do not fit Eq. (9) can be calculated by:

Table 1. YNI model's α and β values [11]

	α_m	α_f	α_h	β_d	β_m	β_p	β_h
C_1	0	0	1.209×10^{-3}	0	40	0	1
C_2	-	-	-6.534	-	-17.8	-	∞
C_3	1	-3.55×10^{-4}	0	-4.21×10^{-3}	0	-2.25×10^{-4}	0
C_4	-1	-1	0	-1	0	-1	1
C_5	-10	5.633	-	2.5	-	13.3	-10
V_0	-37	-20	-20	5	-62	-40	-30

$$\alpha_q = 3.4 \times 10^{-4} \frac{(V+100)}{\exp\left(\frac{V+100}{4.4}\right)-1} + 4.95 \times 10^{-5} \quad (10)$$

$$\beta_q = 5 \times 10^{-4} \frac{(V+40)}{1-\exp\left(-\frac{V+40}{6}\right)} + 8.45 \times 10^{-5} \quad (11)$$

$$\alpha_d = 1.045 \times 10^{-2} \frac{(V+35)}{1-\exp\left(-\frac{V+35}{2.5}\right)} + 3.125 \times 10^{-2} \frac{V}{1-\exp\left(-\frac{V}{4.8}\right)} \quad (12)$$

$$\alpha_p = 9 \times 10^{-3} \frac{1}{1+\exp\left(-\frac{V+3.8}{9.71}\right)} + 6 \times 10^{-4} \quad (13)$$

$$\beta_f = 9.44 \times 10^{-4} \frac{(V+60)}{1-\exp\left(-\frac{V+29.5}{4.16}\right)} \quad (14)$$

Further details regarding the derivation of the equations in this section, and the values of constants mentioned in Table 1, are clearly derived and addressed in [11].

3. The immunity system

The system of immunity is a combination of vital operations, which carried out by particles, cells, and organs distributed throughout the body, to save it from poisons, diseases, alien particles, and harmful cells. This biological system identifies and break down pathogens, like viruses or microbes. The system of immunity can recognize between the alien creatures that cause damage and the healthy body cells, which is considered the most essential function. To stimulate proper mechanisms of protections, alien cells are also categorized. The primary kind of immunity cells is the lymphocytes, which includes two main types of B lymphocytes cells that are produced in the marrow of the bone. The human body contains several different kinds, each one produces 'Y' shape antibodies from their surfaces with a different chemical structure. The second type is T lymphocytes cells, which are created from the thymus, which in turn categorized into T-suppressor cells (TS) and T-helper cells (TH) [13].

Once the antigens invade the organism's body, they are identified by the Antigen Presenting Cell which sends a warning signal to the T cells. Upon receiving the warning, it will stimulate B cells to immediately create antibodies to eliminate those

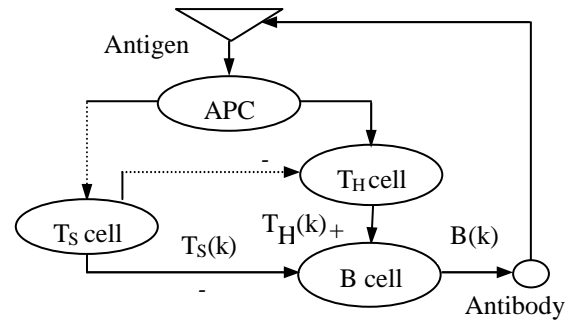


Figure. 2 Mechanism of the immunity feedback control [13]

antigens [14].

Based on the immunity feedback control mechanism shown in Fig. 2, the consistency of the B-cells can be obtained at the kth iteration as follow [13]:

$$B(k) = T_H(k) - T_S(k) \quad (15)$$

$$T_H(k) = K_1 \delta(k) \quad (16)$$

$$T_S(k) = K_2 \{f[\Delta B(k - d)]\} \delta(k) \quad (17)$$

Where the interaction between the antibody that impacts the B-cells and antigen is represented by the nonlinear function $f(\cdot)$, while the consistency of the antigen at the kth generation is represented by $\delta(k)$, $B(k)$ is the B-cell's consistency, the helper gene and the suppressor gene are represented by K_1 and K_2 , and the consistency change of the B-cell is represented by $\Delta B(k - td) = B(k - td) - B(k - td - 1)$, where the immune response time delay is signified by td . The relationship between the antigen and the consistency of B-cells, according to Eq. (15) - (17) can be written as:

$$B(k) = K_1 \{1 - \eta f[\Delta B(k - td)]\} \delta(k) \quad (18)$$

where the proportional coefficient between TS and TH is represented by $\eta = K_2/K_1$.

4. The immune PID controller

The Immune-PID is suggested here as an arrhythmia regulator. The BBO algorithm is used to compute the Immune and PID blocks parameters. The amount of antigen $\delta(k)$ is assumed as the error $e(k)$, and the output of the controller $u(k)$ is the total motivation that is accepted by the B-cells. The feedback control system is represented as:

$$u(k) = K \{1 - \eta f[\Delta u(k - t_d)]\} e(k) = K_I e(k) \quad (19)$$

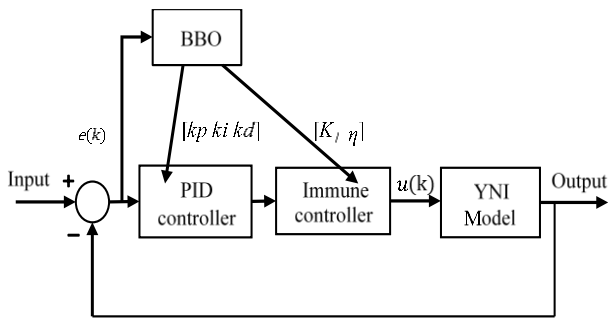


Figure. 3 The block diagram of the Immune-PID controller

Where the proportional gain of immune controller is represented by $K(1 - \eta f[\Delta u(k - td)]) = K_I$, the gene that controls the K_I is $\eta = K_2/K_1$, and $f(\cdot)$ refers to a non-linear function. $f(\cdot)$ is chosen here as the action function of the T-cell's regulator, based on the consistency of the antigen influence on the antibody in the response of immune.

$$f(x) = \frac{1}{1 + \exp(-cx)} \quad , \quad c > 0 \quad (20)$$

Where c has a positive value that determines the active region of x , the range of $f(x)$ is [0-1]. The output of the immune controller can be written as:

$$u(k) = K \left\{ 1 - \eta \left(\frac{1}{1 + \exp(-cx)} \right) \right\} e(k) = K_I e(k) \quad (21)$$

The traditional PID must be combined with the immune controller, because the immune controller is a nonlinear proportional controller. Therefore, it is not robust against the error caused by non-linear disturbance or against noise. The result of this combination can improve the system's performance and overcome this problem as shown in Fig. 3 [13-15].

$$u(k) = k_I \times k_p \left[1 + \frac{k_i}{1-z^{-1}} + k_d(1-z^{-1}) \right] e(k) = k_{IP} \left[1 + \frac{k_i}{1-z^{-1}} + k_d(1-z^{-1}) \right] e(k) \quad (22)$$

where k_{IP} represents the proportional gain.

5. BBO Algorithm

The BBO algorithm is based on the improvement of biogeography, it describes in a mathematical way, the migration of species from one island to another, this migration targets better places to live, the algorithm is proposed in 2008 by D. Simon [16]. The transitions of species from one island to another in biogeography depend on Habitat Suitable Index

(HSI), which in turn depends on several characteristics such as the diversity of food cover, temperature, land area, and water resources. These characteristics are called Suitability-Index-Variables (SIV). HSI is similar to the cost or fitness function in other algorithms of population-based optimization and SIV point out the tuned parameter's number [17]. Generally, two steps are included in the BBO algorithm: information sharing (migrations) and mutation.

5.1 Migrations

The habitat with the highest HSI is the best for living species, therefore, the migration of species between habitats occurs when the HSI of habitats is low, or if more species congregate in the same habitat. The previous process is called Emigration. When the species migrate to a habitat with high HSI and few species, the process is called Immigration. Both Emigration and Immigration are also called migration [18, 19]. The process of migration used to change the solution H_i by involving the properties from another solution H_j , which can be determined via:

$$H_i(SIV_k) \leftarrow H_j(SIV_k) \quad (23)$$

An H_i solution emigration or immigration can be determined according to the emigration rate μ_i and immigration rate λ_i [17]. μ_i and λ_i are calculated using Eq. (24) and Eq. (25), in each habitat which have a different value.

$$\mu_i = \frac{k_i}{n} E \quad (24)$$

$$\lambda_i = \left(1 - \frac{k_i}{n} \right) I \quad (25)$$

Where I and E represent the maximum immigration rate λ and the maximum emigration rate μ respectively, n denotes the population size, k_i is the rank H_i after evaluation.

5.2 Mutations

The probabilistic factor (mutation) is used to add new features to a randomly selected solution via modifying one or more of its SIVs, based on P_i which increases the diversity of population [16], Eq. (26) illustrates the mutation of the i_{th} habitat.

$$M_i = m_{max} \left(1 - \frac{P_i}{P_{max}} \right) \quad (26)$$

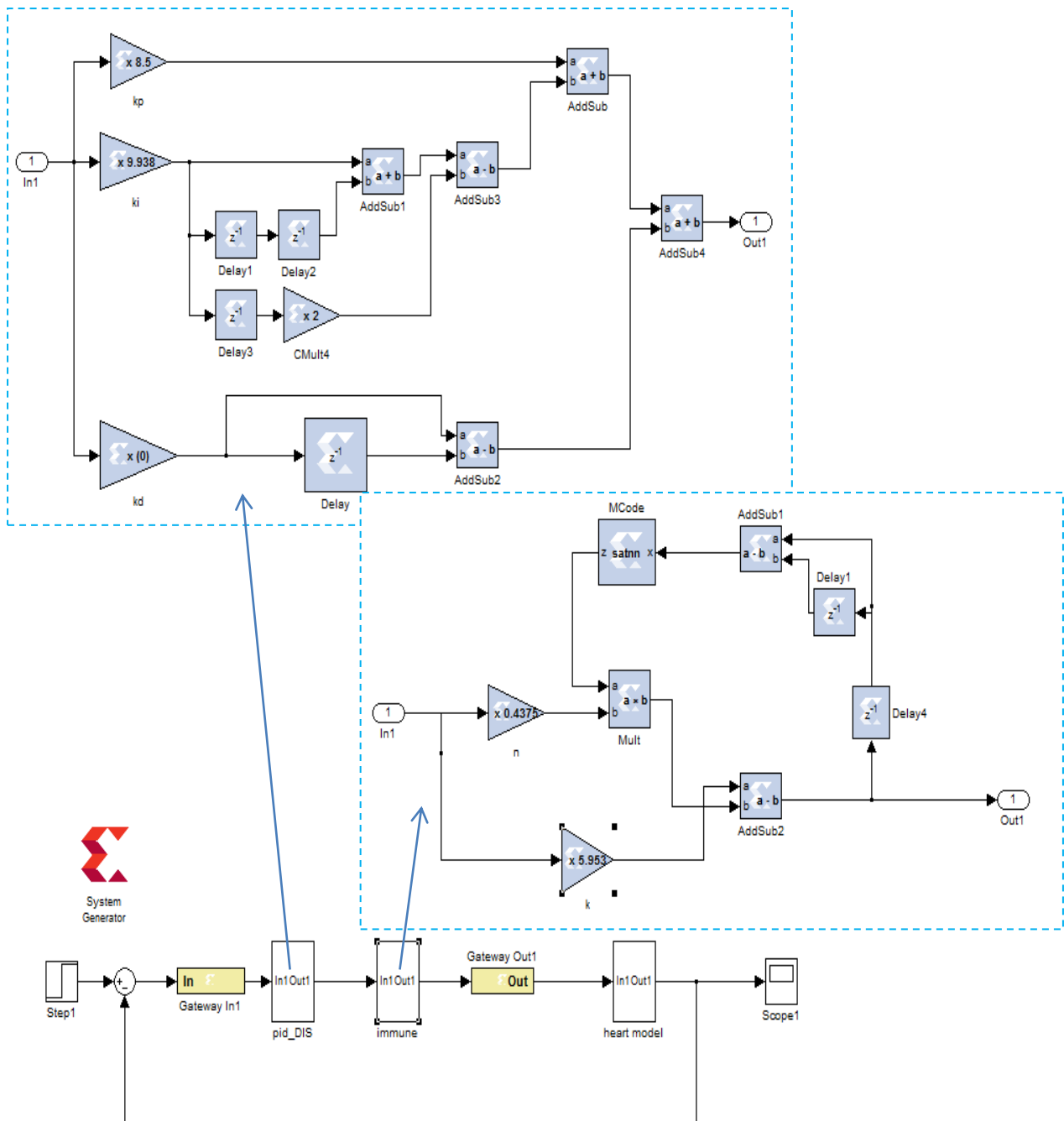


Figure. 4 FPGA design of immune-PID controller

Where P_i is the existence probability of the habitat H_i and m_{max} is defined by the user and represents the mutation coefficient.

6. FPGA implementation

The libraries of the Xilinx System Generator block set [20], are used here to execute the immune-PID controller. In order to implement this controller using FPGA, a Spartan-6 (XC6SLX45T-3FGG484) FPGA Kit is suggested. In this paper the "saturation function" is used instead of the non-linear function $f(x)$ in the immune part of the proposed controller,

due to the difficulty of implementation in FPGA [21]. In order to produce the best resolution, the input signal is represented using fixed-point sign number of 10 bits, with 3 bits binary points (Fix 10 3), while the output signal is represented using (Fix 48 26). Fig. 4 and Fig. 5 show the FPGA design and the device utilization summary of the immune-PID controller with 1% utilization of hardware platform slices respectively. To interface the hardware platform with MATLAB Simulink, JTAG cable is used to configure the suggested controller into the FPGA kit Fig. 6 illustrates this process.

Device Utilization Summary				
Slice Logic Utilization	Used	Available	Utilization	Note(s)
Number of Slice Registers	138	54,576	1%	
Number used as Flip Flops	131			
Number used as Latches	0			
Number used as Latch-thrus	0			
Number used as AND/OR logics	7			
Number of Slice LUTs	351	27,288	1%	
Number used as logic	332	27,288	1%	
Number using O6 output only	269			
Number using O5 output only	32			
Number using O5 and O6	31			
Number used as ROM	0			
Number used as Memory	0	6,408	0%	
Number used exclusively as route-thrus	19			
Number with same-slice register load	8			
Number with same-slice carry load	11			
Number with other load	0			
Number of occupied Slices	127	6,822	1%	
Number of MUXCYs used	304	13,644	2%	
Number of LUT Flip Flop pairs used	393			
Number with an unused Flip Flop	263	393	66%	
Number with an unused LUT	42	393	10%	
Number of fully used LUT-FF pairs	88	393	22%	
Number of unique control sets	1			
Number of slice register sites lost to control set restrictions	5	54,576	1%	

Figure. 5 The device utilization summary of the immune-PID controller

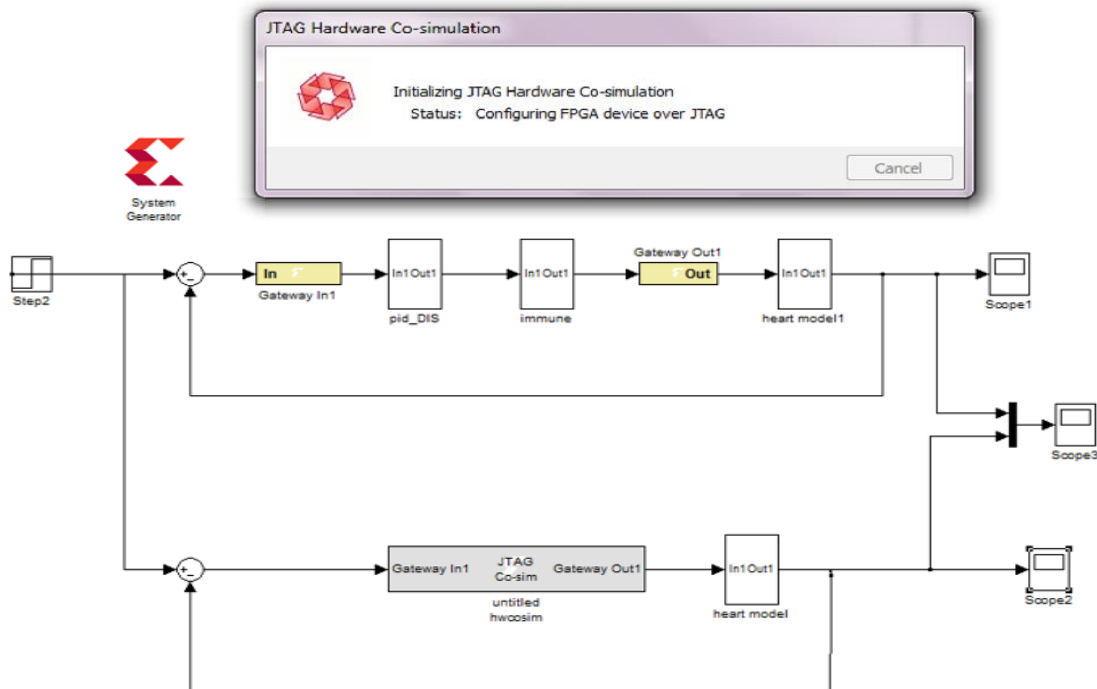


Figure. 6 Downloading the bit stream to the hardware device through JTAG cable

7. Simulation results

The results of simulation for the YNI heart model which addressed in Eq. (1), and the suggested controller based on the BBO algorithm are offered in this section for an HR of 70 bpm. The YNI model response without controller is shown in Fig. 7.

The parameters of the BBO algorithm are considered here as in Table 2.

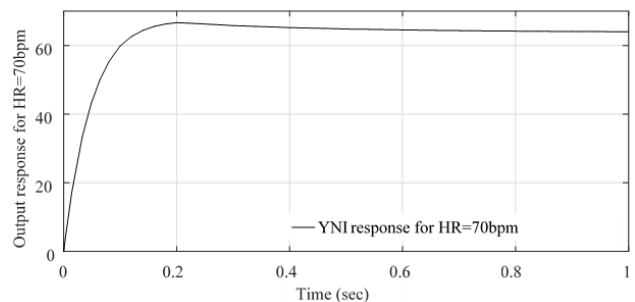


Figure. 7 The response of the YNI heart model for HR= 70 bpm

Table 2. The parameters of the BBO algorithm.

BBO parameter	Value
Generation number	10
Mutation Probability	0.06
n (Habitats)	20
SIV	5
E	1
I	1
Min Domain	0
Max Domain	10

Table 3. Optimal controller parameters.

Parameter	Immune	PID	Immune-PID
k_p	-	3.6085	8.531
k_i	-	6.6866	9.937
k_d	-	0.0027	0.0003
K_1	7.9005	-	5.9601
c	0.5	-	0.5
η	0.1567	-	0.4424

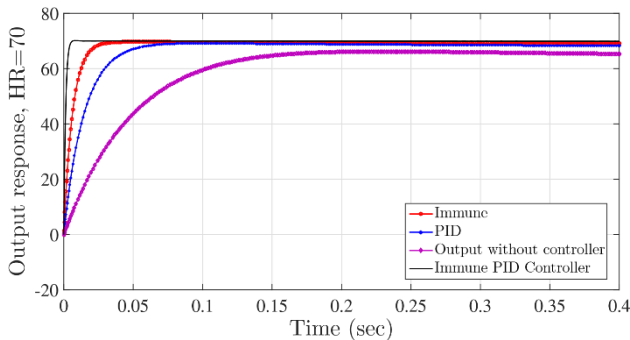


Figure. 8 The YNI model response regulated by immune, PID, and immune-PID controllers for HR=70 bpm

Table 3 illustrates the optimal obtained parameters for the Immune, classical PID, and Immune-PID controllers from the BBO algorithm.

Fig. 8 shows the response of the YNI model after applying the suggested controllers to regulate the HR, using the parameters in Table 3.

It can be noticed from Fig. 8 that the cardiac system with these controllers tracks the desired HR with small rise time (t_r), settling time (t_s), and steady state error ($e.s.s.$).

The comparison between all controllers is shown in Table 4. This table illustrates the performance of the three controllers. The immune-PID has the best average performance with ($M_p = 0\%$, $t_s = 0.01$ sec, $t_r = 0.0023$ sec and $e_{ss} = 0.1$) which satisfies the design requirement. While the other methods of the PID tuned by BBO, the PID and the immune controllers provide a plausible response. Table 4 addresses a direct comparison between the three controllers, and Robust RBF-neural-network based backstepping

Table 4. Optimal controller parameters

Parameter	PID	Immune	Immune-PID	RBF-neural-network[7]
$M_p(\%)$	0	0	0	1.6
$t_r(sec)$	0.0169	0.0135	0.0023	0.155
$t_s(sec)$	0.05	0.04	0.01	0.244
e_{ss}	1.03	0.8	0.1	1.60

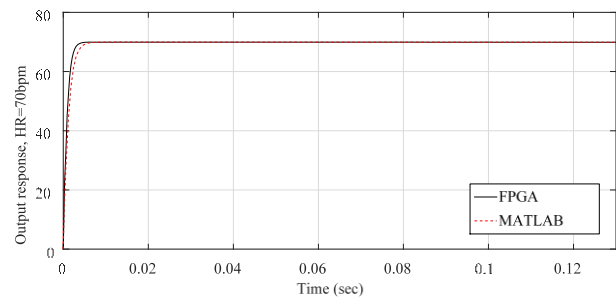


Figure. 9 The YNI model response with HR=70 bpm regulated by immune-PID controller in FPGA

control in [7] with respect to the maximum overshoot (%), rise time, settling time and e_{ss} respectively.

Fig. 9 illustrates the simulation results of the controlled cardiac system in MATLAB and FPGA.

To evaluate the robustness of the proposed controller against random variable input, an HR in the range of (70 ± 10) bpm is considered, then a sudden disturbance is applied to mimic the sudden Tachycardia, which represented by adding step function with amplitude of 40 at the 5th second without removing it. A sudden bradycardia scenario is also evaluated by adding a step function with amplitude of -30 at the 5th second without removing it. The results are similar in both Matlab and FPGA, but little chatter is found in the outperformance of the FPGA caused by the use of a fixed point.

Fig. 10 shows a comparison between the simulated performance of FPGA and Matlab. Where Fig. 10 (a) shows the response of the heart of tracking a random input without external influence on it. While Fig. 10 (b) illustrates its rapid response against Tachycardia, where the proposed controller manages to return to the natural range of HR and overcome this rise within only (0.0185 sec.). The proposed controller is also robust against the sudden Bradycardia and can overcome it and regulate the HR within only (0.0106 sec.) as shown in Fig. 10 (c).

8. Conclusions

In this paper instead of using conventional discrete PID, the Immune-PID controller is suggested for human heart rate regulation. The YNI heart mathematical model is considered here. The

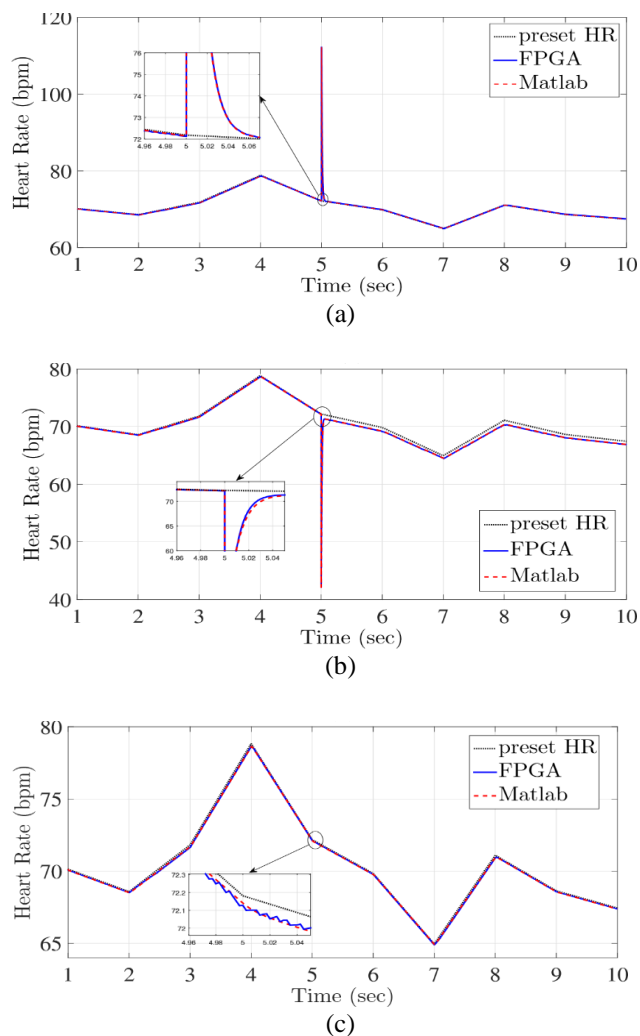


Figure. 10 The HR at rest where: (a) the normal case, (b) with tachycardia, and (c) with bradycardia.

suggested controller optimal parameters are determined using the BBO algorithm. In addition, the optimal Immune-PID controller is implemented using FPGA due to its great data storage capacity, low consumption of energy, and high speed of operation in order to design this controller as a simple Pacemaker circuit. According to the simulation results, the immune-PID controller outperforms the PID, Immune controllers and backstepping controller [7] in terms of t_s , t_r , and e.s.s as shown in Table 4. The proposed controller shows significant robustness in overcoming sudden changes in the HR, for the two scenarios (Tachycardia and Bradycardia), as shown in Fig. 10, the controller successfully returns the HR to its normal range within less than 20 msec. in each case.

Conflicts of Interest

The authors declare no conflict of interest.

Author Contributions

The Algorithm development, methodology, Matlab simulation, FPGA implementation, early draft writing, and results were provided by Eman F. Mohsin. Tariq Tashan and Ekhlash H. Karam jointly directed and supervised the research, reviewed, revised the paper.

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