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Dynamical Indicator of Human Body's Physical Endurance

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Abstract: Physical endurance is the time span between the beginning of physical activity by an individual and the termination because of exhaustion. Physical endurance involve a multifaceted behaviour which can be understood by complexities. Everyone performs physical activity in order to sustain life. However, the amount of activities done are largely subject to personal choice and varies from person to person as well as for a given person over time. Physical activity like meditation/exercises are positively related to physical fitness. One needs to understand relation between physical activity, exercise, physical fitness and health. These activities can be partitioned mutually exclusively into many different ways. This paper categorizes daily physical activity into three broad subdivisions based on amount of body movements taking place are: (i) light, (ii) moderate and (iii) high intensity. These three characterizations are considered to be mutually exclusive and sum up to total energy spent by an individual. The behavior of the three factors physical activity, heart and energy generated is analyzed with the help of Fast Lyapunov indicator (FLI), Dynamic Lyapunov indicator (DLI), Small alignment index (SALI). FLI's increase for chaotic orbits for values of R=20, Q=70 for the case of high intensity exercises and to linearly regular orbits for values of R=5, Q=8 and R=10, Q=12 in the cases of light and moderate exercises respectively. SALI's alters through non-zero value for R=20, Q=70 while it tends to zero for values of R=5, Q=8 and R=10, Q=12. DLI's the largest Eigen values form a definite pattern/curve for n=2000 for values of R=5, Q=8 and n=100 for R=10, Q=12 respectively as the motion stays regular plus dispersed randomly as the motion is chaotic for n=60 and for R=20, Q=70.

Keywords: Physical Endurance, Fast Lyapunov Indicators (FLIs), Dynamic Lyapunov Indicators

(DLIs), Small Alignment Indexes (SALIs), Lyapunov Characteristics Exponents (LCEs)

1. Introduction

Physical activities cause body to utilise more oxygen than resting, exercising, meditating, walking, jogging etc. which enhance and maintain physical fitness, happiness. Also, physical activities are performed for strengthening muscles, cardiovascular system, honing athletic skills, weight loss maintenance, enjoyment. The level of physical fitness varies as amount of physical activity ranging through low to high. Dynamic exercises such as steady running, tend to produce a higher demand of the oxygen during exercise, due to the improved blood flow. On other hand, static exercises such as weight-lifting causes oxygen concentration to rise significantly. In long term, it fills an individual with happiness, refreshes & improves health.

Bhardwaj and Bangia [1] studies complex dynamics of meditating body using nonlinear analysis techniques. Bhardwaj et al [2-3, 8-9] discussed different methodologies and forecasting techniques to predict air pollution, water pollution and weather forecasting with the impact of pollutants on health and the dynamics of environmental and health assessment. The different tools exist to classify regular and chaotic orbits in dynamical systems such that time series curves, phase plots, Poincaré maps, power spectra, etc. These indicators are very powerful, but not sufficient to differentiate regular and chaotic

motion when the system bears higher degrees of freedom. Buckley *et al* [4] discussed about the sedentary office which is a growing case for change towards better health and productivity. Buman and King [5] gave some ideas for exercise as a treatment to enhance sleep. Buskirk and Jeffris [6] discussed chaotic dynamics of coupled nonlinear oscillator. Dinas *et al* [7] studied the effects of exercise and physical activity on depression. The concept of FLI was proposed by Froeschle *et al* [10]. Grassberger and Procaccial [11] measured the strangeness of Strange Attractors. Lega and Froeschle [12] discussed the relationship between Fast Lyapunov Indicator and periodic-orbit for symplectic-mappings. Nonlinear Chaotic Systems discussed by Li *et al* [13]; Martelli [14] and Nagashima and Baba [15]. Nikić [16] discussed importance of yoga. Pimlott [17] gave some idea about miracle drugs. The concept of DLI is first coined by Saha *et al* [18, 19] and SALI by Skokos [20, 21].

In this paper, model consists of compartments of breathing function from physical activities towards energy produced in the body is explained. A mathematical model is developed to study the effects of physical activities on the human body and on its two major components heart and energy generated by body. The model uses non-linear differential equations which are time dependent under constant metabolic-rate. A compartment model of breath function from lungs to tissues or body cells is discussed. Also, it is considered that flow from lungs to the blood depends on the concentration present in the blood since oxygen binds to hemoglobin in the blood. In reality, this is a non-linear relationship but for simplicity, it is considered that it is directly proportional to the difference in concentrations of the two chambers. Firstly, assumptions and factors taken into account to form the non-linear model equations are discussed.

The nature and behavior of various body parts during various forms of exercise postures such as stretching, bending, standing, squatting, etc. are formulated mathematically. As, these factors change with time so, rate of change in heartbeat, lungs and blood flowing through various veins with respect to time are taken into consideration. Thus, the system of nonlinear differential equations defines the model. Phase space analysis of the system of equations is evaluated which involves calculation of equilibrium point, Jacobian matrix defining nature of the point. Further, Eigen-system leads to phase space diagram is discussed. Graphs are plotted for the equilibrium point and phase space. Time series data of the three major dynamic components i.e. Physical state of the body, Heart & Energy in the Body-cells is analyzed. This analysis is used for future predictions. Also, the graphs have been plotted for each component with time. These results are studied and various observations are made which helped in verification and a better understanding of the previously obtained results.

1. Methodology

1.1 Lyapunov Characteristic Exponents (LCE)

LCEs were coined after the Russian Engineer, Alexander M. Lyapunov. Lyapunov exponent measures relative stability of the system of equations. Multifaceted systems having Lyapunov exponents equivalent to the dimensions in the phase-space. Whenever these values are less than zero, implying nearby initial-conditions all converge to one another, and initial small errors decrease with respect to time. However, if any of the LCE value is positive, it shows that nearby initial conditions diverge from one to another exponentially. Thus, the condition is known as sensitive dependence on initial conditions which is coined as **Chaos**. LCE measures the relative stability of the dynamical system(s). Number of LCEs depends on the dimension of phase-space of the system.

1.2 Dynamic Lyapunov Indicator (DLI)

DLI is categorized by the pattern formed by the largest Eigen values. When *J* denotes Jacobian matrix of the dynamical system. At every discretized time, calculate the eigenvalues of the matrix J and then plot the largest eigenvalue. It seems that the eigenvalues of Jacobian, J can be found as:

 $J - \lambda_i I = 0$

where I is the identity matrix and J is Jacobian. Whenever DLI values form a definite pattern, the motion is regular. Otherwise, DLI points get disseminated on a random basis showing no definite pattern, i.e. chaos reaching motion.

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1.3 Fast Lyapunov Indicator (FLI)

FLI represents the value of the largest eigenvector at each iteration and also least dependent on the initial conditions.

Theorem: Consider p-dimensional basis $V_p(0) = (v_1(0), v_2(0),...,v_p(0))$ entrenched into the (general notation) n-dimensional system alongwith initial condition $P(0) = (x_1(0), x_2(0),...,x_n(0))$. It is taken to be the largest among the evolving vector basis outcomes at each iteration.

Thus, it is calculated as:

FLI =supremum $||v_j||, j = 1, 2, ..., m$

where v_i is an eigenvector of eigen-basis having dimension 'm'.

Remark: Deviation norm vector approaches precipitously to totally unalike outcomes for regular vs chaotic orbits differing in huge magnitude indicating discrimination towards regular and chaotic trajectories.

For regular orbits, these are comparatively small values and for randomized orbits, FLI gets very large values. FLI increases exponentially for chaotic-orbits and linearly for regular orbits. FLI indexes present norm value for deviation vector & never computes limiting value, as $t \rightarrow \infty$.

1.4 Small Alignment Index (SALI)

SALI explicitly distinguishes in ordered and chaotic-motion at each time evolution.

Theorem: Assuming 'n' dimensional phase-diagram for the dynamical system having in general the initial condition (I.C.) as: $P(0) = (x_1(0), x_2(0), ..., x_n(0))$. In addition to it, the deviation vector taken to be: $\xi(0) = (dx_1(0), dx_2(0), ..., dx_n(0))$ via point, P(0) simulating alignment indexes resulting in an orbit following evolution of trajectory with time having initial condition P(0) itself and also of two deviation vectors $\xi_1(t), \xi_2(t)$ in the beginning have two different directions.

For each time step, two vectors deviating are: $\xi_1(t)$; $\xi_2(t)$ need to be stabilized. Then, SALI is:

$$SALI(t) = \min \left\{ \left\| \frac{\xi_1(t)}{\||\xi_1(t)||} + \frac{\xi_2(t)}{\||\xi_2(t)||} \right\|, \left\| \frac{\xi_1(t)}{\||\xi_1(t)||} - \frac{\xi_2(t)}{\||\xi_2(t)||} \right\| \right\}$$

For the case of *n*-dimensional system where $n \ge 2$, SALI oscillates through non-zero value for ordered orbits. SALI value tending towards zero in the chaotic trajectories. For 2D mapping, the alignment index tends to zero both for ordered-and chaotic orbits following however, completely different time rates, to distinguish between the two cases.

Attractors: The two types of the attractors that are to be studied are:

1. Equilibrium: $0 > \lambda_1 \ge \lambda_2 \ge \lambda_3 \ge \ldots \ge \lambda_n$;

2. Strange chaotic:
$$\lambda_1 > 0$$
, $\sum_{i=1}^{n} \lambda_i < 0$.

Mathematical Modeling

The mathematical model for three important components: Physical state of the body, Heart and Energy generated in cells/tissues of the body is proposed. The model helps to study the impact of exercises on these factors simultaneously with time and also to study how the change in one component influences the changes in other with respect to time. For the development of model some assumptions that are required to transform the biological condition into a mathematical model are as follows:

• The core assumption considered for modeling the problem states that oxygen comes in and goes out of the lungs during various physical activities like walking, exercising, resting etc. where breathing

function is denoted by R in volume per time units i.e., the concentration of oxygen in the body cells changes with time.

- Depending on the oxygen capacity of lungs, relevant amount of oxygen goes into the heart. Even after this process, there is a possibility that some amount i.e. minimal volume equivalent to 1.2*L* of oxygen remains in the lungs.
- The variation in oxygen levels in the lung during exercising (specifically, exhaling) is also taken into account. It is assumed that the amount of movement from lungs to the heart can be governed by the concentration streamlined into bloodstream; as oxygen fixes to haemoglobin of blood.
- Oxygen drifts from blood towards body cells where energy for performing activities is produced. Lastly, oxygen consumption is proportional to amount obtainable having metabolic rate (*M*).

The Compartmental model is given as follows:



Assuming non-linearity in the system having x that denotes concentration of oxygen flow during any physical state (like exercising, walking, resting) of human body, y representing the concentration of oxygen flow in heart/bloodstream during a particular physical state of body and variable z to represent the concentration of oxygen in body/cells. These three variables fluctuating rates with respect to time are thought of as three new state variables. Let M represents metabolic rate i.e., how much oxygen is utilized in body; R denotes the breath function and for Inhaling rate: R > 0; Exhaling rate: R < 0; η refers to the maximum transfer rate of molecules from the lungs towards bloodstream; f denotes the max transfer rate of molecules through bloodstream to body tissues; the max lung-volume is denoted by V. Let p refers to period of breath, t accounting for breathing period, Q denotes amount of oxygen intake for various conditions of heart (24 L/min). All the concentration are taken to be positive i.e., $x \ge 0$, $y \ge 0$, $z \ge 0$. Therefore, assigning an appropriate dimension to every state variable, following simplified-model is obtained:

$$x'[t] = ax - b(x - y) + RC$$
 (1)

$$y'[t] = x(b-z) - dy$$
 (2)

$$z'[t] = xy - Mz + dy$$
(3)

where, x[t]-concentration of oxygen flow during any physical state (like exercising, walking, resting) of human body, y[t]-concentration of oxygen flow in heart/bloodstream during a particular physical state of body and z[t]-concentration of oxygen in body/cells. Also, $a = \operatorname{sqrt}(R)/\operatorname{Vol}$, $b = \eta$, C=Q, d = f. As per simulations carried out, following included parameters have constant values in equations (1) - (3) to be considered as follows:

Vol=3600 ml, $\eta = 0.1/s$, f = 0.7/s, M = 0.01/s, $Q_{max} = 24L/min$, R = 20s.

R is deliberated from the tidal volume (about 500ml) for the customary breathing function and the maximum lung volume was used and the breath rate is defined as 20s inhale, 20s suspension, and 20s exhale.

3. Results and Discussion

Critical (or more commonly called fixed) points of the system where the changes in the system's behavior/orbits are obtained through equations of system (1), (2) & (3), i.e. x'(t) = 0, y'(t) = 0 and z'(t) = 0 which are

ax - b(x - y) + RC = 0x(b - z) - dy = 0xy - Mz + dy = 0

Jacobian for system of equation (1) - (3) can be given by

$$J = \begin{pmatrix} 1 + a^*k - b^*k & b^*k & 0 \\ b^*k - k^*z[t] & 1 - d^*k & -k^*x[t] \\ k^*y[t] & d^*k + k^*x[t] & 1 - k^*M \end{pmatrix}$$

The Characteristic equation from the Jacobian Matrix is given as: $\lambda^3 + b_1\lambda^2 + b_2\lambda + b_3 = 0$ where,

$$b_{1} = -(3 + ak - bk - dk - kM);$$

$$b_{2} = -(-3 - 2ak + 2bk + 2dk + b^{2}k^{2} + adk^{2} - bdk^{2} + 2kM + ak^{2}M)$$

$$-bk^{2}M - dk^{2}M - dk^{2}x[t] - k^{2}x[t]^{2} - bk^{2}z[t];$$

$$b_{3} = -(1 + ak - bk - dk - b^{2}k^{2} - adk^{2} + bdk^{2} - kM - ak^{2}M + bk^{2}M + dk^{2}M$$

$$+ b^{2}k^{3}M + adk^{3}M - bdk^{3}M + dk^{2}x[t] + adk^{3}x[t] - bdk^{3}x[t] + k^{2}x[t]^{2}$$

$$+ ak^{3}x[t]^{2} - bk^{3}x[t]^{2} - bk^{3}x[t]y[t] + bk^{2}z[t] - bk^{3}Mz[t])$$

Three cases are considered for three different values of R and Q for various physical activities of the body which are explained.

Fig.1 (a), (b), (c), (d) and (e) plots phase diagram, Lyapunov exponents, DLI, FLI and SALI for Case-1. Fig.2 (a), (b), (c), (d) and (e) plots phase diagram, Lyapunov exponents, DLI, FLI and SALI for Case-2. Fig.3. (a), (b), (c), (d) and (e) plots phase diagram, Lyapunov exponents, DLI, FLI and SALI for Case-3.

Phase plot for Case1





Fig.1. (a) Phase-diagram, (b) Lyapunov Exponents, (c) DLI, (d) FLI, (e) SALI for the stable fixed point (Case-1) Phase space for Case 2



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Fig.2. (a) Phase-diagram, (b) Lyapunov Exponents, (c) DLI, (d) FLI, (e) SALI for the stable fixed point (Case-2)



Phase space for Case 3

Fig.3. (a) Phase-diagram, (b) Lyapunov Exponents, (c) DLI, (d) FLI, (e) SALI for the unstable fixed point (Case-3) **Table:1** explains the Routh-Hurwitz Stability for the fixed points computed in the three cases discussed above. **Table:2** shows behaviour of Lyapunov indicators: DLI, SALI and FLI in different intervals for the Case1 (i.e. R = 5, Q = 8).

Table:3 shows the behavior of Lyapunov Indicators for the Case2 (i.e. R=10, Q=12). Table:4 gives the behavior of Lyapunov Indicators for Case 3 (i.e. R=20, Q=70).

Case	Fixed Point	Characteristic	<i>b</i> ₁	<i>b</i> ₃	$b_1b_2 - b_3$	Stability
		Polynomial				
<i>R</i> =5,	(-44.475, 0.00226, 0.09998)	-1746.32	2.189	1746.5(>0)	2.189×1942.8	Stable
<i>Q</i> =8		- 1942.88 <i>n</i>	(>0)		-1746.32 (>0)	
		$-2.1893n^2 - n^3$				
<i>R</i> =10,	(-133.462, 0.000745, 0.0999)	-15931.7	2.189	15931.7	2.1891×17720.1	Stable
<i>Q</i> =12		- 17720.1 <i>n</i>	(>0)	(>0)	-15931.7	
		$-2.1891n^2 - n^3$				
<i>R</i> =20,	(-1557.70, 0.0000635, 0.09999)	2179797.87	2.188	2179797.8	2.188×(-2425340.36)	Unstable
<i>Q</i> =70		– 2425340.36n	(>0)	(>0)	-2179797.87(<0)	
		$+ 2.188n^2 - n^3$				

Table 1: Routh Hurwitz Stability Analysis for the three cases

 Table 2: Lyapunov Dynamic Indicators for Case I: R=5,Q=8

	No. of iterations	Interval	Behaviour
DLI	1000 < <i>n</i> < 2000	DLI values varying 0.99995 - 1	DLI for the case of light exercises
		showing defining pattern.	follow Regular motion.
FLI	10 < n < 30	FLI values increase gradually	FLI for the case of light exercises
		from 2×10^{37} to 12×10^{38} .	follow Regular motion.
SALI	0 < n < 60, the SALI values	Values of SALI varying 0 - 1.4.	SALI for the case of light exercises
	depict great variation in 0-30		follow Regular motion.
	then stabilizing towards zero		
	in 30 to 60.		

Table 3: Lyapunov Dynamic Indicators for Case II: R=10,Q=12

	No. of iterations	Interval	Behaviour
DLI	0 < n < 100	DLI values varies from 1.0070 to	DLI for the case of moderate
		1.0085 shows a definite pattern.	exercises follow follows Regular
			motion.
FLI	0 < n < 800	FLI values increase gradually	FLI for the case of moderate
		from 5×10^5 to 3.5×10^6 .	exercises follows Regular motion.
SALI	0 < n < 50, has greater	Variation observed from 0 - 1.4.	SALI for the case of moderate
	variation in the range 0-20		exercises follows Regular motion.
	but can be observed		
	convergence to zero in the		
	range 20 - 40.		

 Table 4: Lyapunov Dynamic Indicators for Case III: R=20,Q=70

	No. of iterations	Interval	Behaviour
DLI	10 < <i>n</i> < 60	DLI values varies from 0 to 6000 and	DLI for the case of the heavy
		shows a definite pattern till $n = 50$ and	exercises follow Chaotic motion.
		DLI value > 0 but changes to chaotic	
		motion at $n = 50$ and DLI value 4000.	
FLI	10 < n < 60	FLI values increase from 0 to 8 x 10^{18} .	FLI for the case of the heavy
			exercises follow Chaotic motion.
SALI	0 < n < 40 - increases	Chaotic orbits can be concluded in the	SALI for the case of the heavy
	simultaneously but on	range 0 - 1.4.	exercises follow Chaotic motion.
	the whole decreases.		

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Conclusion

A mathematical model for levels of concentration of oxygen during various physical states of the human body like exercising, walking, jogging, resting etc. and two major components heart/ blood stream and the energy produced in the bodycells modeled as non-linearities factored into complex equations taken to be time dependent along with constant M. The stability analysis including the equilibrium points, Jacobian matrix, Eigen-system and its Phase space is estimated. Further, for improving accuracy and model efficiency, time-step analyzations also implemented. It is observed that there is a chaotic behavior in FLI, DLI and SALI graphs for the case where breath function, R and the amount of oxygen intake, Q are the highest, i.e. R=20 and Q=70. The time series graphs of amount of oxygen in physical state (x), amount of oxygen in heart (y) and amount of oxygen in the bodycells (z) variables indicate that on excessive exercising, oxygen supply/intake increases which further increases the oxygen concentration in heart and body cells. FLI's increase for chaotic orbits for values of R = 20, Q = 70 for the case of high intensity exercises and to linearly regular orbits for values of R = 5, Q = 8 and R = 10, Q = 12 in the cases of light and moderate exercises respectively. SALI oscillating around a non-zero value for value of R=20, Q=70while it tends to zero for values of R=5, Q=8 and R=10, Q=12. DLI's the largest Eigen values form a definite pattern/curve for n=2000 for values of R=5, Q=8 and n=100 for R=10, Q=12 respectively as the motion is regular and distributed randomly as the motion is chaotic for n=60 and for R=20, Q=70. It is good to go for some kind of exercise on a daily basis. But, it is recommended to exercise for a limited time and period to avoid any harm to the body. It can be concluded that with various values of breath functions, R and the amount of oxygen intake, Q, concentration of oxygen varies. This means that more exercising leads to more of oxygen intake and further produces more energy which may keep us active, fresh and happy the whole day. In addition, excessive exercise will lead to excessive oxygen which may not be needed in the body and prove to be harmful.

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Conflict of Interest:

Authors declare no conflict of interest.

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