

# Validity of dynamical analysis to characterize heart rate and oxygen consumption during effort tests

D. Mongin, C. Chabert, A. Uribe Caparros, A. Collado, Eric Hermand, O.

Hue, J. R Alvero Cruz, D. S Courvoisier

## ▶ To cite this version:

D. Mongin, C. Chabert, A. Uribe Caparros, A. Collado, Eric Hermand, et al.. Validity of dynamical analysis to characterize heart rate and oxygen consumption during effort tests. Scientific Reports, 2020, 10, pp.12420. 10.1038/s41598-020-69218-1. hal-03187480

# HAL Id: hal-03187480 https://hal.science/hal-03187480v1

Submitted on 2 Apr 2021

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

## Validity of dynamical analysis to characterize heart rate and

# oxygen consumption during effort tests

D. Mongin<sup>\*,+,1,3</sup>, C. Chabert<sup>2,+</sup>, A. Uribe Caparros<sup>1</sup>, A. Collado<sup>2</sup>, E. Hermand<sup>2,4</sup>, O. Hue<sup>2</sup>, J.

R. Alvero Cruz<sup>3</sup>, D. S. Courvoisier<sup>1</sup>

<sup>1</sup> Quality of Care Unit, University of Geneva, Geneva, Switzerland

- <sup>2</sup> ACTES laboratory, UPRES-EA 3596 UFR-STAPS, University of the French West Indies, Guadeloupe, France
- <sup>3</sup> Malaga University, Andalucía Tech., Department of Human physiology, histology, pathological anatomy and physical education, Malaga, Spain
- <sup>4</sup> Research Unit Hypoxie & Poumon, UMR INSERM U1272, University Paris 13 Nord, Paris, France
- <sup>+</sup> Authors contributed equally to this work.
- \* Corresponding author: denis.mongin@unige.ch

## ABSTRACT

Performance is usually assessed by simple indices stemming from cardiac and respiratory data measured during graded exercise test. The goal of this study is to characterize the indices produced by a dynamical analysis of HR and VO2 for different effort test protocols, and to estimate the construct validity of these new dynamical indices by testing their links with their standard counterparts. Therefore, two groups of 32 and 14 athletes from two different cohorts performed two different graded exercise testing before and after a period of training or deconditioning. Heart rate (HR) and oxygen consumption (VO<sub>2</sub>) were measured. The new dynamical indices were the value without effort, the characteristic time and the amplitude (gain) of the HR and VO<sub>2</sub> response to the effort. The gain of HR was moderately to strongly associated with other performance indices, while the gain for VO<sub>2</sub> increased with training and decreased with deconditioning with an effect size slightly higher than VO<sub>2</sub> max. Dynamical analysis performed on the first 2/3 of the effort tests showed

similar patterns than the analysis of the entire effort tests, which could be useful to assess individuals who cannot perform full effort tests. In conclusion, the dynamical analysis of HR and  $VO_2$  obtained during effort test, especially through the estimation of the gain, provides a good characterization of physical performance, robust to less stringent effort test conditions.

## 1 INTRODUCTION

2 Characterization of Heart Rate (HR) and oxygen consumption (VO<sub>2</sub>) related to 3 mechanical power (i.e., speed or power) during standardized graded exercise test (GET) is an 4 unavoidable step in current athlete's performances assessment <sup>1</sup>. These two measurements are 5 also classically used in the scientific field of sport studies as one of the main physiological 6 output to characterize evolution of athlete's performance over time <sup>2–4</sup>.

Current analysis of these parameters is based on two radically different approaches. The first is the use of standard techniques, easily applicable and extensively used. The most common index to characterize the HR recovery is the Heart Resting Rate (HRR) <sup>5</sup>, commonly defined as the difference between HR at the onset of recovery and HR one minute after. This characterization is known to be a good predictor of cardiac problems in medicine <sup>5</sup>, and is an interesting indicator of physical condition and training <sup>6</sup>. The maximum rate of HR increase (rHRI) is a recent indicator showing correlation with fatigue and training in various studies <sup>6</sup>.

This first type of approaches to characterize HR dynamics suffer from two important drawbacks. First, these measurements mix the amplitude of the HR response to effort with its temporal shape. For instance, someone reaching a maximum heart rate of 190 beat/minute and decreasing to 100 beat/min in one minute will have the same HRR as another person reaching 150 beats/minute and decreasing to 60 beats/minute in one minute, although the HR dynamic is different. Secondly and more importantly, they use only a small fraction of the 20 information contained in the entire effort test (e.g., for HRR, the heart rate at the end of 21 exercise and the heart rate one minute later, so two minutes out of a test of 20 to 30 minutes).

22 Regarding standard analysis of respiratory parameters, the main indicators of athlete's 23 performing capacities are the maximal VO<sub>2</sub> reached during the exercise, the maximal aerobic 24 power or the maximal speed reached, and the values of power or speed at the two Ventilatory Thresholds (VTs), corresponding to the lactic apparition (VT1) and the accumulation (VT2) 25 threshold <sup>7</sup>. Although these VO<sub>2</sub> parameters are currently considered among the best indexes 26 of aerobic fitness evaluation<sup>8</sup>, several drawbacks exist. First, determining them requires most 27 28 of the time a visual analysis of the data. Second, they make use of only a part of the gas consumption dynamics, discarding the majority of the information contained in the entire 29 30 effort test.

31 The second approach, based on dynamical system modeling, could allow to more 32 accurately characterize the HR or VO<sub>2</sub> response during effort. Dynamical analysis based on 33 differential equations is an active subject of research in the behavioral field since the seminal work of Boker<sup>9</sup> and has led to numerous studies in the field of psychology and to several 34 methodological advances <sup>10</sup>. First order differential equation approach has the potential 35 ability to adjust HR measurement <sup>11,12</sup> and VO<sub>2</sub> dynamics during variable effort loads <sup>13</sup>. We 36 37 propose here to use a simple first order differential equation coupled with a mixed effect regression to quantify the link between the exercise load during effort test and the resulting 38 39 HR or VO<sub>2</sub> dynamics. Because dynamical models use all the information measured during the effort test, it may allow to accurately assess performance indices using non-maximal 40 41 effort tests.

42 The aim of this study is to characterize the indices produced by the dynamical analysis
43 of HR and VO<sub>2</sub> for different effort test protocols. The construct validity of these new

44 dynamical indices will be provided by testing their links with their standard counterpart. Their ability to detect performance change over two different context of training load will 45 46 determine their predictive validity and sensitivity to change. We will therefore analyze 47 longitudinal data measured for two groups of young athletes with two different protocols. 48 One group should show a performance increase following a three months training period, and 49 the second group should have a performance decrease after an off-season of 6 weeks. The possibility to apply the proposed dynamical analysis to submaximal effort tests will be 50 51 studied by comparing the result of the analysis performed on the full tests with the one 52 performed on only the first part of the test.

- 53 **METHODS**
- 54 Subjects

To test the reliability of the dynamical analysis model, data were acquired in two different populations (Guadeloupe and Spanish athletes) subjected to two different profiles of exercise (step-by-step cycling and continuous intensity running increase) and physiological conditions (training and deconditioning), presented in Table 1.

	Group 1	Group 2	р	
Number of subjects	32	14		
Age (years)	15.06 (1.48)	15.36 (0.84)	0.487	
Weight (kg)	62.54 (11.98)	64.41 (7.29)	0.593	
Height (cm)	172.40 (8.84)	170.91 (4.34)	0.556	
Gender (Male)	19 (59.4%)	14 (100.0%)	0.014	

Table 1 Biometrical data of the two groups studied in this study at baseline.

59

Group 1 consists of 32 young athletes (19 males and 13 females; 15.1±1.5 year-old) of
the Regional Physical and Sports Education Centre (CREPS) of French West Indies
(Guadeloupe, France), belonging to a national division of fencing, or a regional division of

63 sprint kayak and triathlon. GET was performed at the end of the off-competition season, and 64 after 3 months of intense training (3-7 sessions/week). All athletes completed a medical 65 screening questionnaire, and a written informed consent from the participants and the legal 66 guardians was obtained prior to the study. The study was approved by the CREPS Committee 67 of Guadeloupe (Ministry of Youth and Sports) and the CREPS Ethics Committee and 68 performed according to the Declaration of Helsinki.

69 Group 2 consists of 14 young males,  $(15.4 \pm 0.8 \text{ year-old})$  amateur soccer players from 70 Malaga (Spain), performing three weekly training sessions and one weekly competition. A 71 first GET was performed at the end of the soccer season and a second 6 weeks after. All participants were warned to avoid any training activity during this time. The measurements 72 have been used in a previous publication <sup>14</sup> and was approved by the Research Ethics 73 74 Committee of the University of Málaga, Spain (EMEFYDE UMA: 2012-015 report) and 75 carried out according to the principles of the Declaration of Helsinki. Participation in the 76 study was voluntary, and prior to its initiation, written informed consent was obtained from the participants and the legal guardians of those under 18 years of age. 77

78

### Effort test measurement

79 Group 1 performed an incremental testing on an SRM Indoor Trainer electronic 80 cycloergometer (Schoberer Rad Meßtechnik, Jülich, Germany) associated to a Metalyzer 3B 81 gas analyzer system (CORTEX Biophysik GmbH, Leipzig, Germany). The SRM 82 cycloergometer is directly supervised by computer to automatically maintain a constant 83 mechanic workload by adjustment of the brake in accordance to the number of revolutions 84 per minute. Cardiorespiratory parameters were recorded cycle-to-cycle during all the test to 85 obtain HR and VO<sub>2</sub> all along the test session. The effort protocol used consisted of a 3 minute rest phase, followed by a 3 min cycling period at 50 watts, followed by an incremental power 86 87 testing of +15 Watts by minute until exhaustion. Measurements of VO<sub>2</sub>, HR and mechanical

power during the last increment sustained by athletes were respectively considered as  $VO_2$ max, HR max and Maximal Aerobic Power (MAP). At the end of the test, measurements were prolonged during a 3 min period to record the physiological recovery of athletes.

91 Group 2 performed GET on a PowerJog J series treadmill connected to a CPX 92 MedGraphics gas analyzer system (Medical Graphics, St Paul, MN, USA) with cycle-tocycle measurements of respiratory parameters -including VO<sub>2</sub>, and HR- with a 12 lead ECG 93 (Mortara). The stress test consisted of an 8-10 min warm up period of 5 km.h<sup>-1</sup> followed by 94 continuous 1km.h-<sup>1</sup> by minute speed increase until the maximum effort was reached. Power 95 96 developed during the effort test was calculated using the formula described by the American 97 College of Sport Medicine (ACSM). The latter determines an approximate VO<sub>2</sub> of runners<sup>15</sup> 98 associated to the Hawley and Noakes equation that links oxygen consumption to mechanical power<sup>16</sup>. 99

#### 100 Truncated effort tests

101 In order to test the robustness of the dynamical analysis, truncated effort tests were 102 generated from the maximal effort test for both groups. It consisted in removing the 103 measurements of the test for power (or speed) above 2/3 of the maximum power (or 104 maximum speed) value, so that the maximum power (or speed) achieved during the truncated 105 test lies between the two ventilatory thresholds The recovery period was set as the recovery 106 measurements of the full effort test with values below the maximum value reached during the 107 truncated exercise. An example of truncated effort is presented in Fig1, for a VO<sub>2</sub> 108 measurement during an effort test of group 1.



*Fig1:* VO<sub>2</sub> measured during a maximal effort test (light colors lines), and the truncated test generated from these data (dark colors lines)

#### 109 *Standard indices:*

110 The HRR calculated is the standard HRR60, which is the difference between the HR at the onset of the recovery and the HR 60 seconds later. The ventilatory thresholds 1 (VT1) 111 112 and 2 (VT2) are calculated using the Wasserman method using the minute ventilation VE/VO<sub>2</sub> for determining VT1 and VE/VCO<sub>2</sub> for VT2<sup>17</sup>. The rHRI is derived by performing a 113 114 sigmoidal regression of HR before and during the first 3 min effort step (only in group 1) and calculating the maximum derivative from the estimated parameters, as described in <sup>18</sup>. 115 116 Maximum aerobic power is the maximum power spend during the maximal effort test. 117 HRmax and VO<sub>2</sub> max are the maximum values of the rolling mean of HR and VO<sub>2</sub> over 5 118 points.

119

#### New indices using dynamical analysis:

120 A first order differential equation describes a relation between a time dependent 121 variable, its change in time and a possible time dependent excitation mechanism. For a 122 variable *Y* (HR or VO<sub>2</sub>), it reads:

$$\dot{Y}(t) + \frac{Y(t) - Y_0}{\tau} = \frac{K}{\tau} \times P(t) \tag{1}$$

Where  $\dot{Y}(t)$  is the time derivative of Y (i.e. its instantaneous change over time),  $Y_0$  its 123 124 equilibrium value (i.e. its value in the absence of any exterior perturbation) and P(t) the excitation variable, that is the time dependent variable accounting for the exogenous input 125 126 setting the system out of equilibrium. Equation 1 describes the dynamics of a self-regulated 127 system that has a typical exponential response of characteristic time  $\tau$  and an equilibrium value  $Y_0$  in the absence of excitation (i.e. when P(t) = 0). For a constant excitation (i.e. a 128 constant P(t) = P, the system stabilizes at a value KP after several  $\tau$ . This value depends on 129 130 both the system and the excitation amplitude (see Fig2 left panel).



**Fig2** simulated HR dynamics following equation 1, for two different efforts (left panel: constant effort, right panel: effort test of three incremental steps), an equilibrium value of 50 beats.min<sup>-1</sup>, a decay time of 30 s and a gain of 1.

HR and VO<sub>2</sub> are two self-regulated features of our body: they respond to an effort with a certain characteristic time to reach a value corresponding to the energy demand <sup>19</sup>. Equation 1, as already demonstrated in <sup>13</sup> for VO<sub>2</sub>, can reproduce the dynamics of these two measures when considering that P(t) is the power developed by the body during effort.

135	Assuming that HR or $VO_2$ follow equation 1, only three time-independent parameters are
136	needed to characterize and to predict their dynamics for any time dependent effort:
137	- $Y_0$ (i.e. $HR_0$ or $VO_{20}$ ) is the <i>equilibrium value</i> , i.e. the value in the absence of
138	effort.
139	- $\tau$ is the characteristic time or <i>decay time</i> of the evolution of the variable. It
140	corresponds to the time needed to reach 63% of the absolute change of value for a
141	constant excitation. For instance, for an individual running at 10 km/h and who
142	would have a total increase of HR of 60 beats/min for that effort, the decay time
143	would be the time needed to increase his heartbeat by 38 beats/min (60 beats/min
144	× 63%).
145	- <i>K</i> , the <i>gain</i> , is the proportionality coefficient between a given effort increase and
146	the corresponding total HR or VO <sub>2</sub> increase ( $\Delta HR$ and $\Delta VO_2$ ). An illustration is
147	provided in Fig2 left panel: a HR gain of $K_{HR} = 1$ beat/min/W leads to a HR
148	increase of 100 beats/minute for a 100W effort increase, and to $\Delta HR = 200$
149	beats/min for a 200W effort increase.

ing that UD on VO follow equation 1 only three time independent non-motion and

150 An example of the dynamics for HR following equation 1 is given in Fig2 considering  $HR_0 = 50$  beats.min<sup>-1</sup>,  $\tau_{HR} = 30$  s,  $K_{HR} = 1$  and two efforts types. These three coefficients 151 152 tightly characterize the dynamics of HR and allow us to generate the response to any effort.

153 The estimation of the three parameters characterizing the dynamics according to equation 1 is 154 done in a two-step procedure, consisting in first estimating the first derivative of the variable studied over a given number of points with Functional Data Analysis (FDA) regression spline 155 method <sup>10,20</sup>. It consists on generating a B-spline function that fits the outcome to be studied 156 and then estimating the derivative of that function. In order for the generated B-spline 157 158 function to be differentiable, it needs to be smooth. This is achieved through a penalty 159 function controlled by a smoothing parameter. This parameter was chosen to maximize the 160  $R^2$ , which is the goodness of fit of the model to the data.

161 Once the derivative estimated, a multilevel regression is performed to estimate the linear 162 relation between the derivative, the variable and the excitation (summarized by the three 163 parameters presented before).

This two-step estimation procedure has been extensively tested and described in a recent simulation study <sup>12</sup>. It can be applied to data with non-constant time sampling if it contains more than 5 points per typical decay time (in our case, at least one point every 20 seconds) and has a measurement noise below 50% of the signal amplitude.

168 Once the three dynamical parameters estimated, an estimated curve can be reconstructed 169 performing a numerical integration of equation 1 (using the deSolve Package in R<sup>21</sup>).

The above analysis procedure (estimation and estimated curve reconstruction) has been embedded and described in the open-source library *doremi* <sup>22</sup> available in the open source software R. Example code reproducing the analysis presented in this article can be found in the example vignettes associated.

#### 174 Statistical analysis

HR measurements with a rate of change higher than 20 beat.min<sup>-1</sup> from one 175 176 measurement to the next one were first removed as spurious results from the sensors. Indices difference within each group between the first and the second measurement was 177 178 assessed using paired t test, and effect sizes were estimated by Cohen's d index. Associations 179 between standard physical performance indices and the results of our dynamical analysis 180 were assessed using Spearman rank correlation coefficients for continuous variables and 181 logistic regression for dichotomous variables. Training was operationalized as a binary 182 variable set to 0 for measurements before training for group 1 and after deconditioning for group 2 (untrained situation), and to 1 for measurements after training for group 1 and beforedeconditioning for group 2 (trained situation).

185 All analyses were performed using R version  $3.4.2^{23}$ , the package *doremi*<sup>12,22</sup> for the 186 dynamical analysis and the packages *data.table*, *Hmisc* and *ggplot2* for the data management 187 and statistical indicators.

#### 188 **RESULTS**

189 The associations between standard indices were high, especially between the maximum value of oxygen consumption (VO<sub>2</sub> max), the MAP achieved and the ventilatory 190 191 threshold powers for VO<sub>2</sub> (correlations ranging from 0.73 to 0.93). There was also a 192 significant negative correlation between rHRI and VO<sub>2</sub> max (correlation coefficient of -0.42, 193 p = 0.023), meaning that a higher maximum aerobic power reached during effort or a higher 194 maximal VO<sub>2</sub> is associated with a lower rate of HR increase during the first effort test (For 195 full details of these associations, for the first time of measurement, see Supplementary Table 196 1 online).



#### data — estimate — Power (W)

**Fig3** Example of HR and VO<sub>2</sub> dynamics from one subject for each group. Blue line shows the power supplied by the subject during the effort, gray lines are the experimental measurement of HR or VO<sub>2</sub>, and red lines shows the estimation given by the dynamical model.

197

198 Example of HR and VO<sub>2</sub> dynamics is given in Fig3, together with the estimated curve 199 obtained from the first order differential equation analysis. The model was very close to the observed values for both HR and  $VO_2$ , and for both effort test protocols, with  $R^2$  (median 200 201 [IQR]) of 0.96 [0.93, 0.97] for HR, 0.94 [0.92, 0.96] for VO<sub>2</sub> in group 1, and 0.95 [0.91, 0.97] 202 for HR, 0.94 [0.90, 0.96] for VO<sub>2</sub> in group 2. The estimated curve deviate from the 203 experimental data mainly at high effort intensity and at rest before the effort. The ensemble of 204 the estimated values compared to the true observed ones are presented in Supplementary Fig1 205 (online).

The dynamical analysis estimation of resting values overestimated the measured values (HR measures averaged approximately 20 seconds before the first effort increase, see Supplementary Table 2 online), partly because the participants did not provide enough values before the start of the test. Thus, we will discard this index for the rest of the study.

VO<sub>2</sub> max and  $K_{VO_2}$ , the gain of VO<sub>2</sub> (i.e. proportionality coefficient between an effort increase and the final VO<sub>2</sub> increase caused by this supplementary energy expenditure), increased significantly during the 3 months training period of group 1, and decreased significantly during the 6 weeks of deconditioning of group 2 (Table 2). The effect size was slightly higher for  $K_{VO_2}$  than VO<sub>2</sub> max in the two groups and was higher for deconditioning than for training for both variables.

A small decrease of the power of the first ventilatory threshold (power VT1) is also observed in population 2.  $\tau_{VO_2}$ , the response time  $\tau$  of VO<sub>2</sub> to the effort is shorter than  $\tau_{HR}$ , the response time of HR, in both populations. The HR gain ( $K_{HR}$ ) is remarkably similar in both groups, and unaffected by training or detraining. The relative standard deviation of  $\tau_{VO_2}$ (between 35% and 50%) is higher than the relative standard deviation of the associated gain ( $K_{VO_2}$ ). None of the dynamical parameter (gain *K* or decay time  $\tau$ ) displayed significant correlation with the length of the experimental data record.

Measurement Indices Measurement 2 p value Cohen's d 1 **Group 1 : training** MAP(W) 239.8 (55.2) 242.7 (60.0) 0.85 VO<sub>2</sub> max (mL/min/kg) 33.6 (6.1) 42.5 (7.4) < 0.01 1.31 HR max (beat/min) 186.9 (10.1) 188.5 (7.2) 0.51 Power VT1 (W) 97.2 (50.7) 113.9 (40.9) 0.16 Power VT2 (W) 174.7 (54.5) 181.0 (49.8) 0.64 HRR (beat/min) 35.0 (12.3) 36.9 (9.6) 0.53 rHRI (beat/min/s) 0.84 0.5(0.2)0.5 (0.1) 0.82 106.4 (33.7) 108.3 (32.2)  $\tau_{HR}$  (s)  $K_{HR}$  (beat/min/W) 0.43 (0.13) 0.43 (0.12) 0.96 58.9 (19.9) 57.5 (27.7) 0.84  $\tau_{VO_2}$  (s)  $K_{VO_2}$  (mL/min/W) 9.2 (1.1) < 0.01 6.9 (1.7) 1.61 **Group 2 : deconditioning** MAP (W) 231.9 (31.1) 226.4 (27.4) 0.62  $VO_2 \max (mL/min/kg)$ 62.5 (5.6) 49.9 (6.4) < 0.012.11 HR max (beat/min) 199.6 (7.2) 200.6 (5.0) 0.65 Power VT1 (W) 116.3 (20.8) 101.1 (16.9) 0.04 0.80 Power VT2 (W) 184.0 (26.6) 165.2 (29.0) 0.10 HRR (beat/min) 38.3 (9.0) 41.3 (10.3) 0.50 100.2 (37.3) 91.7 (25.1) 0.50  $\tau_{HR}$  (s)  $K_{HR}$  (beat/min/ W) 0.4 (0.1) 0.4 (0.1) 0.64 54.6 (19.4) 58.4 (30.9) 0.70  $\tau_{VO_2}$  (s)  $K_{VO_2}$  (mL/min/W) 12.6 (1.7) 8.9 (1.1) < 0.012.57

**Table 2** Comparison of the classical indices and the indices stemming from the dynamical analysis of VO<sub>2</sub> and HR: the gain K and the decay time  $\tau$ .

Effort test measurements have been performed before and after the 3-month training in group 1 and before and after the 6-week deconditioning in group 2.  $VO_2$ :  $O_2$  consumption; HR: Heart Rate; MAP: Maximal Aerobic Power; HRR: Heart Resting Rate; rHRI: rate of Heart Rate Increase; VT: Ventilatory Threshold. Group 2 does not have rHRI because of the protocol used: the linear increase of power does not allow proper calculation of rHRI. Effect size is given by the Cohen's d of the t test.

223 In univariable analysis,  $\tau_{HR}$  was correlated with measures of HRmax and HRR (Table 3), and  $K_{HR}$  (i.e., proportionality coefficient between effort increase and final HR increase 224 225 caused by this supplementary energy expenditure) was negatively correlated with weight, 226 maximal aerobic power, maximum O<sub>2</sub> consumption, and the two ventilatory thresholds. Only in group 1,  $K_{HR}$  was also negatively correlated with age, height and rHRI, whereas a 227 228 correlation with HRmax is found only in group 2. In other words, a decrease of  $K_{HR}$ , (i.e. a decrease of  $\Delta HR$  for a given effort) was linked with an improvement of oxygen maximal 229 230 consumption, maximal aerobic power and the power corresponding at the two transition 231 thresholds. Overall, correlations with new indices were higher than the correlations found 232 between standard HR indices and other performance variables (see Supplementary Table 1 233 online). In a multivariable analysis performed in each group including age, weight, height, VO<sub>2</sub> max and power at ventilatory thresholds, only weight remained significantly associated 234 235 with  $K_{HR}$  (see Supplementary Table 3 online).

236

Group 1 : training								
	$ au_{HR}$	$K_{HR}$	$\tau_{VO_2}$	$K_{VO_2}$				
Age	0.11	-0.43**	-0.01	0.06				
Weight	0.07	-0.73***	-0.06	0.13				
Height	-0.06	-0.55***	-0.09	0.05				
MAP	0.09	-0.79***	0.12	0.01				
$VO_2$ max	0.07	-0.65***	-0.10	0.57***				
Power VT1	0.13	-0.44***	0.03	0.09				
Power VT2	0.10	-0.63***	-0.03	0.12				
HR max	0.32*	0.16	0.02	0.18				
HRR	-0.52***	-0.14	0.06	0.06				
rHRI	-0.01	0.37**	0.19	-0.24				
Group 2 : deconditioning								
Age	0.07	-0.08	0.10	0.20				
Weight	-0.11	-0.63***	-0.34	0.08				
Height	0.16	0.05	-0.16	-0.13				
MAP	0.28	-0.73***	-0.25	-0.04				
VO <sub>2</sub> max	-0.05	-0.49**	-0.25	0.67***				
Power VT1	-0.05	-0.54**	-0.15	0.19				
Power VT2	0.01	-0.49*	0.10	0.33				
HR max	0.38*	0.49**	0.40*	-0.10				
HRR	-0.72***	0.16	-0.32	-0.08				

**Table 3** Spearman correlation coefficients between the gain K and the decay time  $\tau$  of HR and VO<sub>2</sub> for both populations, physiological characteristics and standard analysis indices.

MAP: Maximal Aerobic Power; HR: Heart Rate; HRR: Heart Resting Rate; VT: Ventilatory Threshold; rHRI: rate of Heart Rate Increase. Significance is indicated as follows: \*: p<0.05; \*\*: p < 0.01, \*\*\*: p < 0.001

VO<sub>2</sub> decay time ( $\tau_{VO_2}$ ) was globally independent of physiological variables and standard indices (Table 3), whereas  $K_{VO_2}$  was strongly associated with VO<sub>2</sub>max. In a multivariable analysis performed on each group including age, weight, height, training, VO<sub>2</sub>max and power at ventilatory thresholds, VO<sub>2</sub>max and training remained significantly associated with  $K_{VO_2}$  (see Supplementary Table 3 online). In group 1, training increased the  $K_{VO_2}$  of 1.1 mL/min/W on average and an increment of 1L/min of VO<sub>2</sub> max increased  $K_{VO_2}$  by 2.7 mL/min/W on average. In group 2, the deconditioning decreased  $K_{VO_2}$  by 2.1 and the decrease of 1L of VO<sub>2</sub> max lowered the VO<sub>2</sub> gain by 1.8 mL/min/W.

### 246 Truncated effort test

When performing the dynamical analysis on the truncated effort tests (see Fig1), the calculated  $R^2$  were slightly lower than the analysis performed on the entire test (0.90 [0.88, 0.94] for VO<sub>2</sub> and 0.93 [0.89, 0.95] for HR in group 1, and 0.90 [0.87, 0.93] for VO<sub>2</sub> and 0.90 [0.87, 0.95] for HR in group 2. The resulting dynamical indices were highly correlated with the one calculated on the entire effort test, as presented in Fig4.



**Fig4** comparison of the dynamical indices estimated on the entire effort test (x axis) and on the truncated effort test (y axis) for  $VO_2$  (top row) and HR (bottom row). The solid black lines represent the identity.

The gain estimated on the truncated effort was slightly higher than the one estimated on the entire effort test. Correlation between the gain (for VO<sub>2</sub> and HR) and the other performance indices remained similar to the ones observed in Table 3. The VO<sub>2</sub> gain  $K_{VO_2}$ estimated on the truncated effort test still significantly changed between the two time points for both groups: from 8.9 (1.6) to 10.2 (1.8) mL/min/W for group 1 (p < 0.01, Cohen's d = 257 0.754), and from 15.0 (2.3) to 12.2 (1.7) ML/min/W for group 2 (p<0.01, Cohen's d = 1.38). 258 In summary, the VO<sub>2</sub> gain presented higher values but still significantly increased with 259 training and decreased with deconditioning.

260

## 261 **DISCUSSION**

#### 262 *Main findings*

263 Modeling the evolution of HR and VO<sub>2</sub> during effort tests with a first order differential equation driven by the power spent during the effort produced an estimation able 264 265 to reproduce in average 95% of the observed variance of HR or VO<sub>2</sub>. The model was 266 successfully tested in two different populations (Guadeloupe and Spanish athletes) subjected to two different profiles of exercise (step-by-step cycling and continuous intensity running 267 268 increase) and physiological conditions (training and deconditioning). The dynamical analysis 269 provided three indices: the equilibrium value or resting state, the decay time, and the gain or 270 proportionality between a given effort increase and the corresponding total increase in HR. 271 HR gain was correlated to the main indices of athlete's performance (MAP, VO<sub>2</sub> max, VT1 272 and VT2), which was not the case of other standard HR indices. Furthermore, VO<sub>2</sub> gain was sensible to change in training or physical deconditioning. Finally, the indices obtained when 273 274 modeling truncated effort test (using about the first 2/3 of the effort tests data) had similar 275 characteristics, showing the robustness and usefulness of such approach to incomplete effort 276 tests. Such incomplete tests could occur due to lack of time but also when assessing older or 277 sick persons.

#### 278 Standard indices

Using standard indices, it was possible to assess the relevance of the training/deconditioning conditions used for this study. Results were in line with those

obtained by other studies <sup>6,19</sup>, thus confirming the quality of the effort tests results in the two 281 282 groups of athletes. In particular, the relationships between ventilatory thresholds (VT), maximal aerobic power (MAP) and maximum oxygen consumption (VO2 max), as well as 283 the change in VO<sub>2</sub> max after 3 months of training and after 6 weeks of deconditioning, were 284 in accordance with expected results <sup>24</sup>. VO<sub>2</sub> max variation was also more pronounced in the 285 deconditioning group than in the training one, as reported in previous observations <sup>25,26</sup>. 286 Concerning rHRI, the negative correlations with VO<sub>2</sub> max and MAP was reported previously 287 288 and is due to a parasympathetic withdrawal with sympathetic activation causing a relatively 289 slower HR increase in response to intensity increase for well-trained athletes when compared to untrained  $^{3,6}$ . 290

291

#### Dynamical analysis

There was a moderate correlation between VO<sub>2</sub> gain ( $K_{VO_2}$ ) and VO<sub>2</sub> max <sup>27</sup>. Under an assumption of linearity between mechanical workload and O<sub>2</sub> consumption, VO<sub>2</sub> max corresponds to the oxygen consumption for the MAP expenditure and is directly linked to  $K_{VO_2}$ :

$$VO_{2max} = VO_{2resting} + MAP \times K_{VO_2}$$
(2)

However, VO<sub>2</sub> max is estimated via a single experimental measurement, 296 supposed to be the  $VO_2$  at the maximum effort achieved by the athletes. The ability to reach 297 298 maximum capacities during effort test is subject to several internal and external factors such 299 as athlete's engagement, mood state, fatigue and many others. Furthermore, the linear relation between energy demand and  $O_2$  consumption may not hold for high power expenditure <sup>28</sup>, 300 301 and thus VO<sub>2</sub> max may not be representative of physical performance for intermediate efforts. In contrast,  $K_{VO_2}$  is estimated from the entire VO<sub>2</sub> dynamics during the effort test, yielding a 302 robust estimate of the VO<sub>2</sub> response to effort. As a consequence, the VO<sub>2</sub> gain estimated on 303

304 truncated effort tests was still sensible to training and deconditioning and seems a promising 305 performance index for submaximal effort test, such as those employed for patients suffering 306 chronic disease or for elderly patients.

The typical response time (i.e. the decay time  $\tau$ ) of VO<sub>2</sub> was shorter than the HR one, in agreement with previous results <sup>29</sup>. This temporal delay of HR compared to VO<sub>2</sub> kinetics is due to the fact that heart flow regulation is partially driven by the oxygen demand of the organism detected via chemoreceptors, causing the HR increase to be a consequence of the VO<sub>2</sub> increase <sup>30</sup>.

312 The high variability of the decay time estimated, may be explained by the different energetic profiles of the athletes according to their sport discipline <sup>31,32</sup> or soccer field 313 position <sup>33</sup> that could modify the kinetic of the VO<sub>2</sub> curve to reach the plateau at each 314 315 increment step of the protocol. The variability of the gain value is the result of the aerobic metabolic efficacy, which is constant according to the substrate<sup>34</sup>, and the cycling or running 316 317 efficacy, which is globally similar in a homogenous population of athletes. These differences may participate to the higher variability of the decay time estimated, compared to the 318 319 variability of the gains.

320 The negative link between the  $K_{HR}$  and subject weight may be explained by the 321 known association between fat-free mass weight and heart's left ventricular size and mass <sup>35</sup>. 322 This association, reflecting a well-trained heart in heavier athletes, results in a lower  $\Delta$ HR for 323 a given effort and so a lower  $K_{HR}$ .

324 Strength and weakness

The main strength of this study is the use of two different populations of athletes, with two different effort tests and two different training schemes, showing its potential generalizability. Nevertheless, further study will need to extend these results to older adults, young children, and people with strong sedentary habits. A second strength is related to the analyses used, which allowed the estimation of performance indices without a maximum effort test. These analyses pave the way to obtaining accurate performance indices and information on training or deconditioning among larger groups of the population, such as the elderly, or patients at risk of cardiovascular events. The availability of ready to use, open source, tools for such analysis should facilitate its use for researchers and sport coaches <sup>22</sup>.

334 As for limitations, the dynamic model used in this study made three assumptions that 335 led to slightly suboptimal fits. First, the assumption that the equilibrium value is constant 336 before and after the effort does not hold and led to the overestimation of these value. Indeed, HR and VO<sub>2</sub> are known to decrease back to their resting value on a longer time scale due to 337 338 the reduction of blood volume (i.e. dehydration), the evacuation of the heat accumulated 339 during the muscular contractions, or the over-activation of the sympathetic system during exercise <sup>36</sup>. The second assumption is that the entire dynamics has one unique characteristic 340 341 exponential time, making the model unable to account for cardiac drift associated to 342 prolonged effort or any long-term modification of the variable dynamics. The third assumption is that the gain of VO<sub>2</sub> and HR is constant along the effort, i.e. that an increase of 343 344 exerted power leads to the same final increase of HR or VO<sub>2</sub>. However, it is known that the VO<sub>2</sub> dynamics saturates at high effort intensities<sup>37</sup> and that the HR response to effort 345 diminishes after the second ventilatory threshold (the inflexion point of the heart rate 346 performance curve<sup>38</sup>). The simple model proposed in this article cannot account for such 347 348 dynamics change, and instead estimates an average gain over the entire effort test. This leads 349 to the increased difference between the estimated curve and the experimental data at high 350 effort intensity. It also explains the higher gains estimated for the truncated effort test, which 351 do not include the part of the effort where the real gain is actually diminishing.

Possibility to release the restrictions listed above is of high interest and is the subject of current research. However, despite the fact that the model can still be improved, it already provides indices with good sensibility to performance change and cardio-respiratory indices used to measure fitness.

356 CONCLUSION

The dynamical analysis of heart rate (HR) and oxygen consumption (VO<sub>2</sub>) during effort appears to be relevant to evaluate performing capacities of athletes and its evolution. It reproduced in average 95% of HR or VO<sub>2</sub> dynamics using only three estimated cardiovascular indices. It was more sensitive to training and deconditioning than classic indices. Furthermore, its ability to extrapolate VO<sub>2</sub> and HR indices from truncated effort tests using only the first steps of the exercise could place it as a valuable tool for evaluate functional capacity from participants unwilling or unable to do maximal exercise testing.

#### REFERENCES

- Beltz, N. M. *et al.* Graded Exercise Testing Protocols for the Determination of VO2max:
   Historical Perspectives, Progress, and Future Considerations. *J Sports Med (Hindawi Publ Corp*) 2016, 3968393 (2016).
- Karvonen, J. & Vuorimaa, T. Heart Rate and Exercise Intensity During Sports Activities.
   *Sports Medicine* 5, 303–311 (1988).
- Bellenger, C. R., Thomson, R. L., Howe, P. R. C., Karavirta, L. & Buckley, J. D.
   Monitoring athletic training status using the maximal rate of heart rate increase. *Journal of Science and Medicine in Sport* 19, 590–595 (2016).
- Smith, T. P., McNaughton, L. R. & Marshall, K. J. Effects of 4-wk training using
   Vmax/Tmax on VO2max and performance in athletes. *Med Sci Sports Exerc* 31, 892–896
   (1999).
- 5. Cole, C. R., Blackstone, E. H., Pashkow, F. J., Snader, C. E. & Lauer, M. S. Heart-Rate
  Recovery Immediately after Exercise as a Predictor of Mortality. *New England Journal of Medicine* 341, 1351–1357 (1999).
- Bellenger, C. R. *et al.* Monitoring Athletic Training Status Through Autonomic Heart
  Rate Regulation: A Systematic Review and Meta-Analysis. *Sports Med* 46, 1461–1486
  (2016).
- 381 7. Binder, R. K. *et al.* Methodological approach to the first and second lactate threshold in
  382 incremental cardiopulmonary exercise testing. *Eur J Cardiovasc Prev Rehabil* 15, 726–
  383 734 (2008).
- 8. Reybrouck, T., Ghesquiere, J., Weymans, M. & Amery, A. Ventilatory threshold
  measurement to evaluate maximal endurance performance. *Int J Sports Med* 7, 26–29
  (1986).

- 387 9. Boker, S. M. & Graham, J. A Dynamical Systems Analysis of Adolescent Substance
  388 Abuse. *Multivariate Behav Res* 33, 479–507 (1998).
- 10. Chow, S.-M., Bendezú, J. J., Cole, P. M. & Ram, N. A Comparison of Two-Stage
  Approaches for Fitting Nonlinear Ordinary Differential Equation Models with Mixed
  Effects. *Multivariate Behav Res* 51, 154–184 (2016).
- 392 11. Zakynthinaki, M. S. Modelling Heart Rate Kinetics. *PLOS ONE* **10**, e0118263 (2015).
- 393 12. Mongin, D. *et al.* Dynamical System Modeling of Self-Regulated Systems Undergoing
  394 Multiple Excitations: First Order Differential Equation Approach. *Multivariate*395 *Behavioral Research* In press (2020).
- 396 13. Artiga Gonzalez, A. *et al.* Kinetic analysis of oxygen dynamics under a variable work
  397 rate. *Hum Mov Sci* 66, 645-658 (2019).
- 398 14. Alvero Cruz, J. R., Ronconi, M., Garcia Romero, J. & Naranjo Orellana, J. Effects of
  399 detraining on breathing pattern and ventilatory efficiency in young soccer players. J
  400 Sports Med Phys Fitness 59, 71–5 (2017).
- 401 15. Ferguson, B. ACSM's Guidelines for Exercise Testing and Prescription 9th Ed. 2014. J
  402 *Can Chiropr Assoc* 58, 328 (2014).
- 403 16. Hawley, J. A. & Noakes, T. D. Peak power output predicts maximal oxygen uptake and
- 404 performance time in trained cyclists. *European Journal of Applied Physiology and*405 *Occupational Physiology* 65, 79–83 (1992).
- 406 17. Wasserman, K., Whipp, B. J., Koyl, S. N. & Beaver, W. L. Anaerobic threshold and
  407 respiratory gas exchange during exercise. *J Appl Physiol* 35, 236–243 (1973).
- 408 18. Nelson, M. J., Thomson, R. L., Rogers, D. K., Howe, P. R. C. & Buckley, J. D. Maximal
- 409 rate of increase in heart rate during the rest-exercise transition tracks reductions in
- 410 exercise performance when training load is increased. *Journal of Science and Medicine in*
- 411 Sport **17**, 129–133 (2014).

- 412 19. Bunc, V., Heller, J. & Leso, J. Kinetics of heart rate responses to exercise. *Journal of*413 Sports Sciences 6, 39–48 (1988).
- 414 20. Trail, J. B. *et al.* Functional Data Analysis for Dynamical System Identification of
  415 Behavioral Processes. *Psychol Methods* 19, 175–187 (2014).
- 416 21. Soetaert, K., Petzoldt, T. & Setzer, R. W. Solving Differential Equations in R: Package
  417 deSolve. *Journal of Statistical Software* 33, 1–25 (2010).
- 418 22. Mongin, D., Uribe, A. & Courvoisier, D. doremi: Dynamics of Return to Equilibrium
  419 During Multiple Inputs. https://CRAN.R-project.org/package=doremi (2019).
- 420 23. R Core Team. R: A Language and Environment for Statistical Computing. (R Foundation
- 421 for Statistical Computing, 2019).
- 422 24. Craig, N. P. *et al.* Aerobic and anaerobic indices contributing to track endurance cycling
  423 performance. *Europ. J. Appl. Physiol.* 67, 150–158 (1993).
- 424 25. Neufer, P. D. The effect of detraining and reduced training on the physiological
  425 adaptations to aerobic exercise training. *Sports Med* 8, 302–320 (1989).
- 426 26. Godfrey, R. J., Ingham, S. A., Pedlar, C. R. & Whyte, G. P. The detraining and retraining
- 427 of an elite rower: a case study. *J Sci Med Sport* **8**, 314–320 (2005).
- 428 27. Arts, F. J. P. & Kuipers, H. The Relation Between Power Output, Oxygen Uptake and
  429 Heart Rate in Male Athletes. *Int J Sports Med* 15, 228–231 (1994).
- 430 28. Zoładź, J. A. & Korzeniewski, B. Physiological background of the change point in VO2
- 431 and the slow component of oxygen uptake kinetics. *J. Physiol. Pharmacol.* 52, 167–184
  432 (2001).
- 433 29. Bearden, S. E. & Moffatt, R. J. VO2 and heart rate kinetics in cycling: transitions from an
  434 elevated baseline. *J. Appl. Physiol.* **90**, 2081–2087 (2001).

- 435 30. Davidson, N. S., Goldner, S. & McCloskey, D. I. Respiratory modulation of barareceptor
- 436 and chemoreceptor reflexes affecting heart rate and cardiac vagal efferent nerve activity.
- 437 *J. Physiol. (Lond.)* **259**, 523–530 (1976).
- 438 31. Gastin, P. B. Energy system interaction and relative contribution during maximal
  439 exercise. *Sports Med* 31, 725–741 (2001).
- 440 32. Barbier, J., Lebiller, E., Ville, N., Rannou-Bekono, F. & Carré, F. Relationships between
- 441 sports-specific characteristics of athlete's heart and maximal oxygen uptake. *Eur J*442 *Cardiovasc Prev Rehabil* 13, 115–121 (2006).
- 33. Wisløff, U., Helgerud, J. & Hoff, J. Strength and endurance of elite soccer players. *Med Sci Sports Exerc* 30, 462–467 (1998).
- 445 34. Frayn, K. N. Calculation of substrate oxidation rates in vivo from gaseous exchange. J
  446 Appl Physiol Respir Environ Exerc Physiol 55, 628–634 (1983).
- 35. Whalley, G. A. *et al.* Association of fat-free mass and training status with left ventricular
  size and mass in endurance-trained athletes. *J. Am. Coll. Cardiol.* 44, 892–896 (2004).
- 449 36. Wyss, C. R., Brengelmann, G. L., Johnson, J. M., Rowell, L. B. & Niederberger, M.
- 450 Control of skin blood flow, sweating, and heart rate: role of skin vs. core temperature.
  451 *Journal of Applied Physiology* **36**, 726–733 (1974).
- 452 37. Yoon, B.-K., Kravitz, L. & Robergs, R. V'O2max, Protocol Duration, and the V'O2
  453 Plateau. *Medicine & Science in Sports & Exercise* 39, 1186–1192 (2007).
- 454 38. Hofmann, P. *et al.* Heart rate performance curve during incremental cycle ergometer
  455 exercise in healthy young male subjects. *Medicine & Science in Sports & Exercise* 29,
  456 762 (1997).