

Heart rate does not accurately predict metabolic intensity during variable intensity roller-skiing or cycling

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2 Abstract

- 3 *Purpose:* To examine the utility of heart rate (HR) and power output (PO) to predict metabolic
- 4 rate (MR) and oxygen consumption (VO₂) during variable intensity roller-skiing and cycling.
- 5 *Methods:* National-level cyclists (n=8) and cross-country skiers (n=9) completed a preliminary
- 6 session to determine VO_{2max} , and a variable intensity protocol with three high-intensity (HI)
- 7 stages at 90% VO_{2max} for 3-min interspersed with three moderate-intensity (MI) stages at 70%
- 8 VO_{2max} for 6-min. Cardiorespiratory measures were recorded throughout. Linear regressions 9 between MR and VO₂ with HR and PO were computed from the preliminary session for all
- 10 athletes and used to predict MR and VO_2 from both HR and PO, separately, during the variable
- 11 intensity protocol. Mean differences with 95% limits of agreement (LOA) between measured
- 12 and predicted MR and VO_2 during the variable intensity protocol were calculated. *Results:* MR
- and VO_2 estimated from HR displayed an overall mean bias close to 0 but wide LOA. HR overestimated MR and VO_2 during MI but underestimated MR and VO_2 during HI, for both
- roller-skiing and cycling. MR and VO_2 estimated from PO was more consistent across the time
- 16 of the experimental trial, displaying a mean bias further from 0 but with tighter LOA.
- 17 **Conclusions:** This study has demonstrated that HR has limited utility to predict metabolic
- intensity during variable intensity roller-skiing and cycling because of wide LOAs. On the other
- 19 hand, metabolic intensity predicted from PO had tighter LOAs, suggesting better reliability. PO
- 20 might provide a better prediction of metabolic intensity compared to HR.
- 21
- 22 Keywords: energy expenditure, intermittent exercise, metabolic rate, power output, VO₂

Review

1 Introduction

The basis of heart rate (HR) as a measure of internal exercise intensity is rooted in the 2 assumption of a linear relationship with oxygen consumption (VO_2) and metabolic rate (MR) 3 during steady-state, sub-maximal intensity exercise,¹ where research has shown nearly perfect 4 5 correlation coefficients (r = 0.99).² Therefore, HR monitoring has been promoted as a valid measure of internal exercise intensity during aerobic steady-state exercise, but not during 6 7 intermittent activity or exercise which involves significant contributions from anaerobic energy systems.³ For example, previous research has reported that HR can provide an accurate means 8 of MR and total energy expenditure prediction over extended durations in free-living 9 conditions⁴ and during continuous steady-state exercise.⁵ However, HR was poor at estimating 10 VO_2 during intermittent exercise, where predicted VO_2 from HR underestimated VO_2 by as 11 much as 10% VO_{2max} during competition and training in handball.⁶ Inaccuracies of up to 10% 12 VO_{2max} could influence interpretation of data and exercise prescription. 13

14 Despite this limitation, HR monitoring might have utility to measure average internal exercise intensity because overestimation and underestimation of exercise intensity during intermittent 15 activity likely evens out. This enables the comparison of the average relative exercise intensity 16 17 between athletes with varying levels of physical capacities and might provide a means to quantify exercise volume (e.g., Banister's TRIMP).⁷ Information regarding average exercise 18 intensity and exercise volume provide valuable information for coaches but has limited utility 19 to tailor competition-replicating training programs to achieve optimal performance 20 improvements. 21

All taken together, HR can provide a useful measure to predict internal exercise intensity during 22 steady-state exercise. However, there is limited evidence for the validity of HR to measure 23 24 intensity during intermittent exercise. Furthermore, cardiovascular drift means that HR increases gradually during prolonged exercise, probably as a result of a declining stroke 25 volume.⁸ Accordingly, this further impacts the associations between HR, VO₂, MR and external 26 27 intensity, such as power output (PO). Despite these limitations, measurement of HR for monitoring exercise intensity remains common place in many intermittent and endurance 28 sports.^{9,10} For example, researchers have proposed calculating time in HR training zones for 29 monitoring elite endurance athletes,¹¹ with some of the proposed zones being as tight as just 30 4% HR_{max}.^{11,12} Misclassification of the training 'zone' when monitoring the athlete might lead 31 to errors in calculating training demands and lead to either over- or under-training and/or 32 33 increased injury risk.

Considering that most competitive sports are intermittent it is of great importance to assess the accuracy of HR to predict VO_2 and MR in these conditions. Therefore, this study aimed to examine the utility of HR to predict VO_2 and MR during variable intensity roller-skiing and cycling. A secondary aim was the compare the utility of HR and PO to predict VO_2 and MR during the same conditions.

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40 Methods

41 Participants

Eight male national-level cyclists (age: 26 ± 4 years; stature: 183 ± 7 cm; body mass: 75 ± 9

43 kg; $VO_{2max} 61 \pm 4 \text{ mL} \cdot \text{kg} \cdot \text{min}^{-1}$) and nine male national-level cross-country skiers (age: 31 ± 6 44 years; stature: $183 \pm 4 \text{ cm}$; body mass: $78 \pm 7 \text{ kg}$; $VO_{2max} 70 \pm 5 \text{ mL} \cdot \text{kg} \cdot \text{min}^{-1}$) were recruited

45 for participation in this study. These participants represent Tier 4 and Tier 5 level athletes

according to the sports participant classification framework.¹³ The participants in this study

represented cyclists who were members of the highest-ranked team within their nation and 47 48 skiers who were either currently or previously competing at the International Ski Federation world cup or the Olympics. All participants provided written informed consent and completed 49 all requirements of the study. Data collected for this research was part of a larger project, some 50 of which has been previously published (XXXX references intentionally withheld for blind 51 review). The regional ethical review board in (XXXX intentionally withheld) (registration 52 XXXX intentionally withheld) preapproved the research techniques 53 number: and experimental protocol. All research was conducted in accordance with the Code of Ethics of 54 the World Medical Association (Declaration of Helsinki). 55

56 Design

The present study consisted of two separate testing sessions: 1) a preliminary testing session which involved an incremental exercise protocol for the determination of cardiorespiratory responses during submaximal and maximal exercise; and 2) a variable intensity experimental protocol with continuous recordings of cardiorespiratory measures, completed within 2 weeks of the preliminary test

61 of the preliminary test.

All testing was completed in laboratory-controlled conditions in a manner as described previously (XXXX_reference_intentionally_withheld_for_blind_review). Briefly, participants were asked to consume a controlled diet. Respiratory variables were measured using an ergospirometry system (AMIS 2001 model C, Innovision A/S, Odense, Denmark), based on the mixed expired method using an inspiratory flowmeter with a sampling frequency of 0.1 Hz. Heart rate was measured continuously using a Polar S610 monitor (Polar Electro Oy, Kempele, Finland). Blood samples (20 mL) were used to determine lactate concentration (Bla) using a

- Biosen 5140 (EKF-Diagnostic GmbH, Magdeburg, Germany).
- All cycling was performed on an SRM high-performance bicycle ergometer (Schoberer Rad
 Messtechnik, Julich, Germany) PO during the cycling was calculated from the calibrated strain
 gauge fixed to the crank arms of the cycle ergometer.

All skiing was performed on roller-skis, using the diagonal stride technique on a motor-driven treadmill (Rodby RL 3000, Rodby Innovation AB, Vänge, Sweden). The PO during rollerskiing was calculated using an equation as previously described by Andersson et al.¹⁴ Briefly, the sum of the power exerted to elevate the total mass (m_{tot} ; body+equipment) against gravity (a) and to every amorganize registering ($u_{tot} = 0.02$) as previously described by Ainegreen et al.¹⁵).

- (g) and to overcome rolling resistance ($\mu_R = .022$ as previously described by Ainegren et al.¹⁵).
- 78
- 79 Preliminary Test
- 80 *Cyclists*

To establish VO_{2max} and the PO for the variable intensity protocol, cyclists performed an incremental test, which was continued to volitional fatigue. The incremental test was performed according to previously described methods Padilla et al.¹⁶ with a modification of the start intensity to 85 W. Each stage of the test was 4-min long with 35 W increments interspersed with 1 min periods performed at 50 W. Cyclists were told to keep their cadence between 80–90 rev·min⁻¹ throughout the test. The test was performed to exhaustion or terminated if the cadence fell below 70 rev·min⁻¹. Maximal PO attained was determined as previously described.¹⁷

88 Skiers

Two preliminary tests were performed in order to establish maximum oxygen consumption (VO_{2max}) and the speed required during the variable intensity experimental protocol. For the

first test, the initial exercise stage was set at an incline of 3° and a treadmill velocity of 8.5 km·h⁻¹. Thereafter exercise intensity was increased by 1° and 0.5 km·h⁻¹ at every 4-minute increment. Skiers completed five to seven submaximal increments interspersed with 1-minute rest for capillary blood lactate sampling. The mean of the three highest consecutive VO₂ values within the last minute of each submaximal increment was used to calculate the speed and inclination necessary to define the percentage of VO_{2max} during the variable intensity experimental test protocol.

Following a 10-min rest period, an incremental protocol for determination of VO_{2max} was performed, starting at an incline of 4° with a progression of 1° each minute with the initial speed set between 10 km ·h⁻¹ and 11 km ·h⁻¹. If participants were able to continue beyond an incline of

101 11°, speed was increased by 0.3 km \cdot h⁻¹ every 30 s.

102 Data Analysis

Data from the preliminary testing session is presented in Table 1. Metabolic rate was calculated 103 using the Weir equation, as previously described¹⁸ and expressed in Joules per second (i.e., 104 Watts [W]). Linear regressions between HR and MR, HR and VO₂, PO and MR as well as PO 105 and VO₂ were computed for all athletes for all exercise stages except the final two. The final 106 107 two stages were excluded from the linear regressions in order to remove the plateau of physiological variables as athletes approached VO_{2max} , ensuring that only the linear portion of 108 the relationships were included. The coefficient of determination for the regressions between 109 HR and MR were nearly perfect for both skiers ($R^2 = 0.99 \pm 0.01$) and for cyclists ($R^2 = 0.98 \pm$ 110 0.01). The coefficient of determination for the regressions between PO and MR were nearly 111 perfect for both skiers ($R^2 = 1.00 \pm 0.00$) and for cyclists ($r^2 = 1.00 \pm 0.00$). The coefficient of 112 determination for the regressions between HR and VO₂ were nearly perfect for both skiers (R^2 113 = 0.99 ± 0.00) and for cyclists (R² = 0.99 ± 0.00). The coefficient of determination for the 114 regressions between PO and VO₂ for skiers ($R^2 = 0.99 \pm 0.00$) and cyclists ($R^2 = 0.99 \pm 0.00$) 115 were also nearly perfect. Individual linear relationships were then used to predict MR and VO₂ 116 117 from both HR and PO separately, permitting the comparison between measured MR and 118 measured VO₂ to predicted MR and predicted VO₂ during the variable intensity exercise trial.

119

Table 1

120 *Experimental Test*

The experimental test began with a 10-min warm up period at an intensity corresponding to 121 50% of VO_{2max} as determined during the preliminary test. Thereafter, the athletes performed a 122 test protocol consisting of three high intensity (HI) stages corresponding with 90% of VO_{2max} 123 for 3-min each interspersed by three moderate intensity (MI) stages corresponding with 70% 124 VO_{2max} for 6-min each (Figure 1). Cardiorespiratory variables were measured throughout the 125 variable intensity protocol as described above for the incremental test, with the mean during the 126 final minute of each stage used for analyses. For skiers, the speed and incline during the test 127 128 protocol were established from the preliminary submaximal test so as to correspond to the two relative exercise intensities (90% and 70% of VO_{2max}). For Cyclists, the power outputs 129 corresponding with 90% and 70% VO_{2max} were computed from the preliminary test from the 130 linear regression equations. 131

132

Figure 1

133 Statistical Analyses

All statistical analyses were performed using IBM SPSS Statistics for Windows (Version 27.0; IBM Corporation, NY) with level of significance set at $\alpha < 0.05$. Shapiro-Wilk tests confirmed

136 that the assumption of normality was not violated and group data were expressed as mean \pm standard deviation (SD). A repeated measure mixed model analysis of variance (within factor: 137 intensity; between factor: exercise mode) was used to determine if exercise mode (cycle vs. 138 skiing) influenced the pattern of physiological response to intensity (HI vs. MI) throughout the 139 variable intensity protocol. For all ANOVAs, effect sizes are presented as partial eta-squared 140 statistic (η^2_p) . Significant interactions were followed up with simple main effect analyses with 141 pairwise comparisons using Bonferroni correction. Further, mean differences between 142 measured MR and predicted MR as well as measured %VO_{2max} and predicted %VO_{2max} from 143 linear regressions with HR or PO during the experimental trial with 95% limits of agreement 144

145 (LOA) were determined according to methods described by Bland and Altman.¹⁹

146

147 **Results**

The power output and measured physiological variables during the experimental trial are displayed in Table 2. There was no interaction effect for any variable ($F_{1,15} \le 3.303$, $p \ge 0.089$, $\eta^2_p \le 0.180$), suggesting that the physiological response was similar between skiing and cycling.

151 As expected, PO, MR, VO_2 , RER, HR, and Bla were greater for HI compared to MI (p < 0.05).

152

The difference between measured and predicted MR and %VO_{2max} from linear relationships 153 with HR and PO during the MI and HI exercise bouts for both cycling and skiing are displayed 154 in Figure 2. For both cycling and roller-skiing, HR underestimated measured MR and %VO_{2max} 155 in the first HI stage and overestimated measured MR and %VO_{2max} in the final MI stage. During 156 cycling, predicted MR and %VO_{2max} from PO were consistently underestimated throughout the 157 variable intensity trial, except for the first stage where MR was overestimated by just 0.6 ± 29 158 W. For roller-skiing, predicted MR from PO overestimated measured MR by 35 ± 76 W in the 159 first HI stage and underestimated measured MR by 77 ± 66 W in the final MI stage, or $1.8 \pm$ 160 2.3% VO_{2max} overestimated in the first HI stage and underestimated measured VO₂ by 4.7 \pm 161 2.6% VO_{2max}. 162

163

Figure 2

Figure 3 displays Bland and Altman plots demonstrating the agreement between measured MR 164 and measured %VO_{2max} with predicted MR and predicted %VO_{2max} from linear relationships 165 with HR and PO. The mean bias for MR predicted from HR was -4 W (95% LOA: -234 to 226 166 W). The mean bias for VO₂ predicted from HR was -0.2 mL·kg·min⁻¹ (95% LOA: -7.8 to 7.3 167 mL·kg·min⁻¹) or -0.4 % VO_{2max} (95% LOA: -12.1 to 11.3% VO_{2max}). The mean bias for MR 168 predicted from PO was -33 W (95% LOA: -163 to 98 W). The mean bias for VO₂ predicted 169 from PO was -1.4 mL·kg·min⁻¹ (95% LOA: -6.1 to 3.3 mL·kg·min⁻¹) or -2.1 % VO_{2max} (95% 170 LOA: -9.3 to 5.1% VO_{2max}). 171

172

Figure 3

More specifically, the mean bias and 95% LOA for each specific exercise mode and intensity is presented in Table 3 and Table 4. MR predicted from HR displayed a mean bias ranging from -107 - 102 W often with wide LOA. MR predicted from PO displayed a mean bias ranging from -70 - 21 W, with tighter LOA.

177 *** Table 3 here ***

*** Table 4 here ***

6

- 178 VO₂ predicted from HR displayed a mean bias ranging from -5.9 4.7% VO_{2max} often with 179 wide LOA. VO₂ predicted from PO displayed a mean bias ranging from -4.8 - 1.5% VO_{2max} 180 with tighter LOA.
- 181

182

183 **Discussion**

This study aimed to examine the utility of HR to predict MR and VO₂ during variable intensity 184 roller-skiing and cycling. A secondary aim was to compare the accuracy of HR and PO in the 185 prediction of MR and VO₂ during the same exercise conditions. The main findings from this 186 187 study were: 1) MR and VO₂ estimated from HR displayed an overall mean bias close to 0 but with wide LOA; 2) more specifically, HR tended to overestimate MR and VO₂ during MI 188 exercise but underestimate MR and VO₂ during HI exercise, for both roller-skiing and cycling; 189 3) MR and VO₂ estimated from PO was more consistent across the time of the experimental 190 trial, displaying a mean bias further from 0 but with tighter LOA. 191

192 The HR-VO₂ relationship is commonly utilised in sports science for exercise prescription and monitoring. For example, HR was first proposed as a means of monitoring exercise intensity 193 during steady-state submaximal exercise because of this relationship.^{20,21} The results from this 194 study have confirmed that nearly perfect regressions exist between HR and MR/VO₂ during 195 laboratory controlled sub-maximal incremental roller-skiing and cycling exercise ($R^2 \ge 0.98$). 196 However, during the variable intensity exercise protocol there was variability in the difference 197 between measured and predicted MR/VO₂ based on this relationship. This suggests that, 198 although nearly perfect regressions exist in sub-maximal steady-state conditions, HR is less 199 accurate at predicting MR and VO₂ during variable intensity exercise, which is common in 200 201 many sports and competitions.^{6,22,23}

202 The results from the present study might suggest that average HR displays good validity but poor reliability to estimate MR and VO₂ during variable intensity cycling and roller-skiing. This 203 is because the overall HR displayed a mean bias close to 0 for both MR and %VO_{2max} but with 204 wide LOA. However, when investigating the mean bias and LOA for individual intensities and 205 exercise modes (Table 3 and Table 4), HR tended to overestimate both MR and VO₂ during MI, 206 but underestimate MR and VO₂ during HI. Accordingly, the under- and overestimation 207 208 throughout the variable intensity exercise trial evened out, resulting in a mean bias close to zero. The time-course change in the accuracy of HR to predict MR or VO₂ might be related to 209 cardiovascular drift.⁸ Accordingly, during endurance exercise training or competitions (e.g., 210 distance cycling or skiing), it is reasonable to expect that HR will become less accurate over 211 time to predict MR and VO_2 . This provides further weighting to the notion that HR at any given 212 point in time during variable intensity exercise does not necessarily reflect the corresponding 213 metabolic intensity.²⁴ This is potentially problematic for athletes who are following a training 214 program where intensity is monitored based upon their HR. For example, at the beginning of a 215 variable intensity training session, the actual metabolic intensity required to reach a given HR 216 might be well above that intended. On the other hand, by the end of an extended training session, 217 HR might severely misrepresent the metabolic intensity. Accordingly, this could lead to 218 misclassification of the training 'zone' when monitoring the athlete causing errors in calculating 219 220 training demands and lead to either over- or under-training and/or increased injury risk and underperformance. 221

222 Quantifying the duration of exercise that athletes perform in various 'HR zones' is common 223 practice.¹¹ In addition, a common measure of so-called '*training load*' (or exercise volume) is

the summated HR zones model, also known as 'Edward's TRIMP',25 which is the 224 amalgamation of exercise duration and an intensity weighting factor based upon HR intensity 225 zones. The results from the present study suggest that the error associated with HR means that 226 with just a 4% HR_{max} zone, the associated MR and VO₂ could be markedly different. This error 227 is in addition to the known day-to-day variation in the HR responses to exercise,^{21,26} as well as 228 the error associated with the actual HR measuring device itself.^{27,28} Such differences, even 229 small, could still have an impact on the prediction of MR and VO₂ and therefore the actual 230 metabolic intensity of exercise. Further, hydration status and environmental factors, such as 231 temperature and altitude can also influence HR.²¹ It is recognised that HR monitoring is 232 undoubtedly a practical tool to provide an indication of exercise intensity. However, all taken 233 together, it seems clear that HR is not necessarily able to reflect metabolic intensity with 234 accuracy at any point in time during variable intensity exercise. Due to variability in the 235 accuracy of HR to predict MR or VO₂, practitioners are unable to be certain of the actual 236 metabolic intensity associated with the HR response. This brings into question the utility of HR 237 intensity 'zones' for exercise prescription and monitoring during variable intensity exercise. 238

Conversely, in the present study, PO had poor validity but good reliability to estimate MR and %VO_{2max} in the same conditions. In demonstration of this, both MR and %VO_{2max} predicted from PO displayed a mean bias further from 0, but with tighter LOA. Explanatory factors for the overestimated MR and %VO_{2max} during the three MI stages of cycling and roller-skiing might be related to an increased metabolic cost due to blood lactate clearance.²⁹ Given that there were tight LOAs, a correction factor could possibly be used to allow a more valid estimate of VO₂ or MR from PO during roller-skiing and cycling.

246

247 Practical Application

The results from this study suggest that PO might provide a better prediction of metabolic 248 intensity during variable intensity cycling and skiing compared to HR. Practitioners could 249 250 calculate individual relationships between PO and MR/VO₂ during laboratory testing sessions. Subsequently, these relationships can be used to predict metabolic intensity during training 251 sessions, likely with greater accuracy compared with using a HR monitor. Accordingly, the 252 training demands of elite-level cyclists and skiers can be better monitored compared with using 253 HR alone. It should be considered that measuring PO from cycling is a common and simple 254 task given modern power metres can be installed onto the crank.^{30,31} Although computing PO 255 during rolling-skiing on a treadmill is also relatively simple, calculating PO during outdoor 256 rolling-skiing or on-snow skiing is a complicated task given variations in snow and weather 257 conditions, as well as variations in air resistance from day-to-day. However, recently some 258 researchers have proposed novel methods of measuring propulsive power during outdoor roller-259 skiing³² and on-snow skiing^{33,34} using inertial sensors. Future research should assess the utility 260 of these models of measuring PO to predict MR and VO₂ during ecologically valid settings. 261

262

263 Conclusion

This study has demonstrated that HR has limited utility to predict metabolic intensity during variable intensity roller-skiing and cycling. On average, MR and $%VO_{2max}$ predicted from HR had a low mean bias but wide LOA. More specifically, HR tended to overestimate both MR and $%VO_{2max}$ during MI exercise but underestimate during HI exercise. As such, HR can provide a good estimate of the average metabolic exercise intensity during variable intensity exercise. However, this brings into question the utility of using HR and 'HR zones' for

270 prescribing and monitoring exercise of variable intensity. Misclassification of the training 271 'zone' when monitoring the athlete causing errors in calculating training demands and lead to 272 either over- or under-training and/or increased injury risk and underperformance. On the other 273 hand, MR and $%VO_{2max}$ predicted from PO had a mean bias further from 0, but with tighter 274 LOA, suggesting better reliability to predict metabolic intensity.

275

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- 280

281 Statements and Declarations

- 282 The authors declare no financial interests or conflicts of interest.
- 283

284 Author Contributions

- GB conceived and designed the research. GB conducted all experiments and processed all data.
- 286 CS analysed data, performed all statistical tests and wrote the manuscript with editorial
- assistance from GB, EA and KS. All authors approved the final version of the manuscript.

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372 Figure Captions

373 Fig. 1 Schematic of the research timeline including preliminary and experimental exercise trials.

374 Numbers represent exercise intensity as a percentage of VO_{2max}.

375

- 376 Fig. 2 Differences between measured and predicted metabolic rate (top row) and %VO_{2max} (bottom
- row) from linear relationships with heart rate or power output for cyclists (left), skiers (right).
- 378 Mean ± SD. %VO_{2max} = Percentage of maximum oxygen consumption; MI = Moderate Intensity; HI =
 379 High Intensity.

380

- **Fig. 3** Bland and Altman plots with 95% limits of agreement (LOA) for measured metabolic rate (top
- row) and measured %VO_{2max} (bottom row) and MR and %VO_{2max} predicted from heart rate (left) and
- 383 power output (right).
- 384 %VO_{2max} = Percentage of maximum oxygen consumption; MI = Moderate Intensity; HI = High Intensity.

Review

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Fig. 1 Schematic of the research timeline including preliminary and experimental exercise trials. Numbers represent exercise intensity as a percentage of VO2max.

108x60mm (300 x 300 DPI)



Fig. 2 Differences between measured and predicted metabolic rate (top row) and %VO2max (bottom row) from linear relationships with heart rate or power output for cyclists (left), skiers (right).
 Mean ± SD. %VO2max = Percentage of maximum oxygen consumption; MI = Moderate Intensity; HI = High Intensity.

197x163mm (300 x 300 DPI)



Fig. 3 Bland and Altman plots with 95% limits of agreement (LOA) for measured metabolic rate (top row) and measured %VO2max (bottom row) and MR and %VO2max predicted from heart rate (left) and power output (right).

%VO2max = Percentage of maximum oxygen consumption; MI = Moderate Intensity; HI = High Intensity.

210x153mm (300 x 300 DPI)

Stage	VO₂ (mL·kg·min⁻¹)	ΫO ₂ (% of ΫO _{2max})	HR (beats∙min⁻¹)	Metabolic Rate (W)	Power Output (W)
			Skiers		
1	32 ± 3	47 ± 4	118 ± 16	872 ± 122	159 ± 26
2	38 ± 4	57 ± 5	131 ± 17	1045 ± 123	200 ± 30
3	46 ± 4	69 ± 6	148 ± 15	1256 ± 159	248 ± 34
4	53 ± 3	80 ± 6	163 ± 14	1474 ± 156	300 ± 38
5	61 ± 4	91 ± 6	175 ± 11	1701 ± 182	379 ± 61
6	69 ± 5	99 ± 2	186 ± 11	1973 ± 168	487 ± 58
7	75 ± 3	100 ± 0	177 ± 3	2169 ± 2	547 ± 12 (n = 2)
			Cyclists		
1	21 ± 2	35 ± 2	106 ± 11	554 ± 43	85
2	26 ± 2	42 ± 4	116 ± 11	669 ± 41	120
3	30 ± 3	49 ± 5	127 ± 11	785 ± 61	155
4	35 ± 4	57 ± 5	138 ± 12	916 ± 33	190
5	40 ± 4	65 ± 5	148 ± 13	1041 ± 39	225
6	44 ± 5	72 ± 6	160 ± 10	1166 ± 50	260
7	50 ± 5	80 ± 7	170 ± 8	1305 ± 55	295
8	54 ± 5	88 ± 7	179 ± 7	1434 ± 48	330
9	58 ± 5	94 ± 5	183 ± 7	1552 ± 119	365
10	59 ± 4	98 ± 3	191 ± 5	1700 ± 95	400 (n = 6)
11	62 ± 3	100 ± 0	195 ± 5	1828 ± 17	435 (n = 4)
12	61 ± 0	99 ± 1	196 ± 3	1793 ± 17	470 (n = 2)

Table 1. Oxygen consumption, heart rate, metabolic rate and power output for skiers and cyclists during the preliminary tests.

Mean \pm SD. HR: Heart rate; R²: Coefficient of determination; $\forall O_2$: oxygen consumption.

Table 2. Power output and physiological variables for cycle and ski exercise during high intensity (HI) and moderate intensity (MI) exercise bouts.

	Power Ou	utput (W)	Metabolio	Rate (W)	Measure	d %VO _{2max}	R	ER	%НГ	R _{Max}	Lactate (n	nmol·L ⁻¹)
	Cycle	Ski	Cycle	Ski	Cycle	Ski	Cycle	Ski	Cycle	Ski	Cycle	Ski
HI1	341 ± 32	366 ± 45	1475 ± 135	1708 ± 168	89.9 ± 2.4	89.1 ± 2.1	1.05 ± 0.06	1.02 ± 0.04	83 ± 5	83 ± 5	2.8 ± 1.3	5.1 ± 1.5
MI1	253 ± 25*	264 ± 35*	1192 ± 122*	1390 ± 141*	74.5 ± 3.3*	74.1 ± 3.2*	0.94 ± 0.02*	0.92 ± 0.02*	80 ± 7*	78 ± 7*	3.3 ± 1.6*	4.1 ± 1.1*
HI2	341 ± 32	366 ± 45	1502 ± 188	1735 ± 188	91.0 ± 2.5	90.1 ± 2.7	1.07 ± 0.03	1.04 ± 0.03	89 ± 6	88 ± 5	5.1 ± 1.9	4.7 ± 0.9
MI2	253 ± 25*	264 ± 35*	1226 ± 120*	1370 ± 131*	76.2 ± 3.2*	72.7 ± 2.8*	0.96 ± 0.03*	0.95± 0.03*	85 ± 6*	82 ± 7*	3.9 ± 2.1*	3.9 ± 1.5*
HI3	341 ± 32	366 ± 45	1518 ± 132	1724 ± 188	91.4 ± 2.6	88.9 ± 2.6	1.11 ± 0.04	1.07 ± 0.04	93 ± 5	90 ± 5	5.9 ± 2.3	5.3 ± 1.4
MI3	253 ± 25*	264 ± 35*	1217 ± 117*	1391 ± 134*	75.4 ± 3.2*	73.7 ± 2.9*	0.97 ± 0.03*	0.96 ± 0.03*	87 ± 6*	85 ± 7*	5.0 ± 2.7*	5.0 ± 2.7*

Mean ± SD. * = Different to HI (p < 0.05). HI: High intensity; MI: Moderate Intensity; VO_{2max}: Maximum oxygen consumption; HR: Heart rate; RER: Respiratory exchange ratio. Table 3. Mean bias and 95% limits of agreement (LOA) for metabolic rate predicted from heart rate and power output during cycle and ski exercise for the high intensity (HI) and moderate intensity (MI) exercise bouts.

	Н	R	Power		
	н	MI	н	МІ	
	Mean Bias (95%	Mean Bias (95%	Mean Bias (95%	Mean Bias (95%	
	LOA)	LOA)	LOA)	LOA)	
Cycle	-107 (-268 to 54)	79 (-5 to 162)	-22 (-79 to 34)	-61 (-165 to 43)	
Ski	-91 (-223 to 40)	102 (-51 to 256)	21 (-123 to 165)	-70 (-195 to 56)	

Table 4. Mean bias and 95% limits of agreement (LOA) for %VO_{2max} predicted from heart rate and power output during cycle and ski exercise for the high intensity (HI) and moderate intensity (MI) exercise bouts.

	н	R	Power		
	н	МІ	н	MI	
	Mean Bias (95% LOA)	Mean Bias (95% LOA)	Mean Bias (95% LOA)	Mean Bias (95% LOA)	
Cycle	-5.9 (-16.2 to 4.3)	3.5 (-0.9 to 7.8)	-0.9 (-4.4 to 2.6)	-4.8 (-11.4 to 1.8)	
Ski	-4.0 (-10.8 to 2.7)	4.7 (-3.1 to 12.5)	1.5 (-3.3 to 6.3)	-4.5 (-9.6 to 0.5)	

HI: High intensity; MI: Moderate Intensity; LOA: Limits of agreement; HR: Heart rate

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