Exercise tolerance during flat over-ground intermittent running: modelling the expenditure and reconstitution kinetics of work done above critical power

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Abstract
Purpose  We compared a new locomotor-specific model to track the expenditure and reconstitution of work done above critical power (W’) and balance of W’ (W’B AL) by modelling flat over-ground power during exhaustive intermittent running.

Method  Nine male participants completed a ramp test, 3-min all-out test and the 30–15 intermittent fitness test (30–15 IFT), and performed a severe-intensity constant work-rate trial (SCWR) at the maximum oxygen uptake velocity (\(\dot{V}O_2\max\)). Four intermittent trials followed: 60-s at \(\dot{V}O_2\max + 50\% \Delta_1\) (\(\Delta_1 = \dot{V}O_2\max - \text{critical velocity } [V_{\text{Crit}}]\)) interspersed by 30-s in light (SL; 40% \(\dot{V}O_2\max\)), moderate (SM; 90% gas-exchange threshold velocity [V\(\text{GET}\)]), heavy (SH; V\(\text{GET}\) + 50% \(\Delta_2\) [\(\Delta_2 = V_{\text{Crit}} - V_{\text{GET}}\)]), or severe (SS; \(\dot{V}O_2\max - 50\% \Delta_1\)) domains. Data from Global Positioning Systems were derived to model over-ground power. The difference between critical and recovery power (DCP), time constant for reconstitution of W’ (\(\tau_{\text{W'}}\)), time to limit of tolerance (TLIM), and W’B AL from the integral (W’B ALint), differential (W’B ALdiff), and locomotor-specific (OG-W’B AL) methods were compared.

Results  The relationship between \(\tau_{\text{W'}}\) and DCP was exponential (\(r^2 = 0.52\)). The \(\tau_{\text{W'}}\) for SL, SM, and SH trials were 119 ± 32-s, 190 ± 45-s, and 336 ± 77-s, respectively. Actual TLIM in the 30–15 IFT (968 ± 117-s) compared closely to TLIM predicted by OG-W’B AL (929 ± 94-s, \(P > 0.100\)) and W’B ALdiff (938 ± 84-s, \(P > 0.100\)) but not to W’B ALint (848 ± 91-s, \(P = 0.001\)).

Conclusion  The OG-W’B AL accurately tracked W’ kinetics during intermittent running to exhaustion on flat surfaces.

Keywords  Critical power · Exercise tolerance · Mechanical modelling · Over-ground power

Abbreviations
ANOVA  Analysis of variance
COM  Centre of mass
CP  Critical power; highest sustainable power output without progressive loss of homeostasis
CV  Coefficient of variation
d  Cohen’s \(d\)
DCP  Difference between recovery power and CP
GET  Gas-exchange threshold
GPS  Global positioning system
ICC  Intraclass correlation coefficient
OG-W’B AL  Calculation of over-ground W’B AL derived from a locomotor-specific regression equation

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Introduction

The curvilinear relationship between athletic performance and time was originally described by Hill (1925), where constant power output maintained to the limit of tolerance (\(T_{\text{lim}}\)) declined as a function of exercise duration. The asymptote of the hyperbolic power–duration relationship has since been termed critical power (CP; critical metabolic rate; associated external power output measured in watts [W]), while the curvature constant represents the finite work capacity above CP, termed \(W'\) (measured in kilojoules [kJ]) (Monod and Scherrer 1965). This relationship holds across various species (Lauderdale and Hinchcliff 1999; Billat et al. 2004) and, among humans, extends to a variety of locomotive modalities, including over-ground running. Here, the terms critical velocity (\(V_{\text{crit}}\)) measured in \(\text{m}\cdot\text{s}^{-1}\) and \(D\) measured in metres (m) are substituted for the external power output associated with CP and \(W'\), respectively. Whilst the CP or \(V_{\text{crit}}\) is typically measured over several days and bouts of constant load exercise, it has been shown that the finite work capacity above CP (\(W'\)) can be completely utilized in a single all-out, three-min exercise test (3 MT) (Vanhatalo et al. 2007). This permits the reliable and valid calculation of an equivalent CP and a \(W'\) value—the work end power (WEP) (Vanhatalo et al. 2007; Wright et al. 2017). The 3 MT has also been applied to over-ground running (Pettit et al. 2012). Therefore, this single-visit test permits quantification of work done above and below the CP or \(V_{\text{crit}}\); hence, the two-parameter model. Parameters derived from the power–time relationship can be used to describe a ‘gold standard’ demarcation of the metabolic steady state (CP; Jones et al. 2019) and the finite work capacity of individuals’ > CP (\(W'\)), which can be used in combination to determine exercise performance (Jones et al. 2010; Jones and Vanhatalo 2017).

Parameters of the power–duration relationship have been incorporated into a composite mathematical framework, designed to estimate the limits of tolerance within the severe-intensity domain during constant work-rate exercise (Monod and Scherrer 1965). Morton and Billat (2004) later applied the two-parameter CP model to intermittent exercise, on the premise that power output during work and rest would be above and below CP, respectively (i.e., when power output is > CP, \(W'\) is expended; when power output < CP, \(W'\) is being reconstituted). However, among other limitations, this model assumed linear depletion and repletion of \(W'\), which simplifies the behaviour of exercise and recovery energetics (Ferguson et al. 2010). In an attempt to address this limitation, Skiba et al. (2012) modelled \(W'\) kinetics during intermittent cycling exercise using an integral equation, where the balance of \(W'\) (\(W'_{\text{BAL}}\)) could be determined from the instantaneous difference between recovery power output and CP, termed the \(D_{\text{CP}}\). This model assumes that \(W'\) reconstitution follows a predictable exponential time course, according to a time constant for \(W'\) reconstitution (\(\tau_{\text{w'}}\)) and accepts that the recovery \(\tau_{\text{w'}}\) varies as a curvilinear function of \(D_{\text{CP}}\). These modifications seem logical, based on the finding of slower PCR recovery kinetics at the end vs. start of intermittent exercise (Chidnok et al. 2013b). This model was successfully applied to a competitive cyclist, using retrospective data obtained from a power meter. Near to complete utilisation of \(W'_{\text{BAL}}\) coincided with \(T_{\text{lim}}\) and subsequent termination of the race (Skiba et al. 2012). More recently, a differential method for the dynamic tracking of \(W'_{\text{BAL}}\) has been developed to overcome the inherent limitations with a mode-specific \(\tau_{\text{w'}}\) for cycling (Skiba et al. 2015). The differential method negates the need for fitting a continuous time function for the dynamic tracking of \(W'_{\text{BAL}}\) using a \(\tau_{\text{w'}}\) based on the independently measured \(W'\) (from the 3 MT) divided by a known \(D_{\text{CP}}\). Both integrative and differential modelling of \(W'_{\text{BAL}}\) offer potential insights into the limitation of intermittent exercise; however, neither model has been applied to whole-body exercise other than cycling, such as over-ground running. This is important, since it is unknown whether current \(W'_{\text{BAL}}\) models can account for the large changes in mechanical power output during exercise and recovery that would be anticipated during whole-body dynamic exercise.

Characterisation of \(W'\) and CP is uncommon among intermittent team-sport athletes, despite its direct relevance to the competitive demands of training and competition, where
frequent surges into the severe-intensity domain are interspersed with periods of lower intensity recovery (Jones and Vanhatalo 2017). With the advent of micro-technology, such as Global Positioning Systems (GPS), over-ground speed can be readily measured in real time (Cummins et al. 2013). Furthermore, estimations of whole-body mechanical work done during over-ground running can be determined (Furlan et al. 2015; Gray et al. 2018). Herein, it is important to distinguish the conversion of metabolic energy to mechanical work between exercise modes through assigned metabolic efficiencies, such as that between cycling (~ 0.25–0.30) and over-ground running (~ 0.50–0.60) (Cavagna and Kaneko 1977). Over-ground running mandates utilisation of elastic recoil in musculotendinous structures, leading to greater corresponding efficiency values for a given energetic input, thus differentiating derived mechanical work and estimated external power outputs (Zamparo et al. 2019). The model developed by Gray et al. (2018) algebraically sums positive and negative external work done across body segments, primarily based on running velocity (i.e., from GPS), alongside known participant characteristics and environmental conditions. Based on the above assumptions regarding mechanical efficiency, both mechanical and metabolic power can be determined. Estimation of over-ground external power output using the above model, coupled with known independently determined parameters of the power–duration relationship (CP and $W'$), should theoretically permit dynamic modelling of $W_{\text{BAL}}$ during running-based exercise. Therefore, the aim of the current study was to develop a locomotor-specific $W_{\text{BAL}}$ model (OG-$W_{\text{BAL}}$) to predict $T_{\text{LIM}}$ during severe-intensity intermittent over-ground running to exhaustion, among well-trained intermittent team sports players. It was hypothesised that the OG-$W_{\text{BAL}}$ model would more closely predict $T_{\text{LIM}}$ than the previously established integral equation, and that it could be used interchangeably with the differential model.

### Materials and methods

#### Participants

Nine healthy males (mean ± SD: age 23 ± 4 years; body mass 77.8 ± 5.5 kg; stature 175.8 ± 5.5 cm; $VO_{2\text{max}}$ 51.1 ± 5.3 mL·kg$^{-1}$·min$^{-1}$) representing university and semi-professional teams (football $n = 7$, rugby union $n = 1$, field hockey $n = 1$) provided written informed consent to take part in this study. A-priori sample size estimation was calculated using G*Power software (Version 3.1.9.3). This was estimated according to $\tau_{wp}$ modified by Bartram et al. (2018), who reported a negative bias of 112 (± 46-s) compared to $\tau_{wp}$ calculated from the differential method. Calculations revealed that eight participants would yield a power (1-beta) of 0.81 at $\alpha = 0.05$. All participants had actively competed in team sport ≥ 3 years. Participants were instructed to refrain from strenuous exercise and avoid alcohol consumption during the 24-h preceding each trial. On the day of testing, participants were also asked to abstain from caffeine intake and arrive at least 3-h postprandial in a euhydrated state. The study received approval from St Mary’s University ethics committee (ref: SMEC_2018-19_056).

### Experimental overview

Participants visited on nine occasions, with each trial separated by at least 48-h. All testing was conducted on a 400-m outdoor synthetic track at a similar time of day (± 3-h). Ambient temperature, relative humidity, and wind speed ranged between 9 and 20 °C, 44 and 87%, and 4.8 and 22.4 km·h$^{-1}$, respectively. For each protocol, a 10-Hz GPS device (FieldWiz, ASI, Lausanne, Switzerland) was fitted between the participant’s shoulder blades and secured to the body within a harness to restrict movement artefacts. The FieldWiz GPS device has provided comparable (CV = 2.0–5.6%; ICC = > 0.8) and reliable (CV = 0.8–2.2%; ICC = > 0.9) measures of peak velocity and total distance during linear and multidirectional motion (Willmott et al. 2019) in relation to a previously validated device (Varley et al. 2012). During the main trials to exhaustion, running velocity was regulated by a pre-recorded audio cue that corresponded to cones placed 10 m apart measured around the 400-m track using a trundle wheel (Voche®, Glasgow, UK). The audio cues were subsequently projected through a portable amplifier (Block Rocker Sport, ION, Cumberland, USA). This allowed for auto-regulation of running velocity by the participant and verification of the investigator. For intermittent velocities, audio cues were edited, time aligned, and looped using commercially available software (Audacity® 2.3.0, USA).

### Experimental protocols

#### Determination of $vVO_{2\text{max}}$ and velocity at gas-exchange threshold

The first visit comprised an incremental ramp test (Vam-Eval) conducted on a 400 m outdoor synthetic track. It commenced at 8.0 km h$^{-1}$ and increased by 0.5 km h$^{-1}$ every min thereafter. The Vam-Eval, as previously implemented by Buchheit et al. (2012), is a modified version of the validated University of Montreal Track Test (Léger and Boucher 1980), with auditory signals matched to cones placed at 20-m intervals. This test was selected to reduce variability of pacing between intervals. End-stage running velocities are also strongly related to $VO_{2\text{max}}$ ($r = 0.96$) (Léger and Boucher 1980). Testing was terminated upon volitional
exhaustion or an inability to sustain the required velocity for two consecutive 20 m intervals. Breath-by-breath pulmonary gas exchange was measured throughout the Vam-Eval using a COSMED K4b² metabolic gas analyser (COSMED, Rome, Italy). Before each test, calibration procedures were performed, requiring ambient room air calibration of gas fractions and against known compositions (16.00% O₂ and 5.00% CO₂). The flowmeter was calibrated using a 3-L volume syringe. Data were averaged every 30 s and aligned to the centre of each time interval (i.e., 0.25, 0.75, 1.25 min, etc.) in line with the previous recommendations (Robergs et al. 2010). Errant breaths (e.g., coughs, swallows, etc.) > 4 standard deviations from the mean were removed (Lamarra et al. 1987). The highest 30-s mean VO₂ was taken as VO₂max. The average velocity during the final 30 s of the Vam-Eval, as derived from GPS data, was taken as vVO₂max (km h⁻¹). Velocity at the gas-exchange threshold (VGET) was verified using the following methods: (1) V-slope method (VCO₂ vs. VO₂) (Beaver et al. 1986); and (2) an increase in minute ventilation (Ve) relative to VO₂ but no increase relative to VCO₂ (Vₑ/VO₂ vs. Vₑ/VCO₂) (Caiozzo et al. 1982). Two-thirds of the ramp rate was deducted from the calculated VGET and vVO₂max to account for mean time response of VO₂ during ramp protocols (Whipp et al. 1981). Strong verbal encouragement was provided throughout the test.

3-min all-out exercise test (3 MT)

Visits 2 and 3 comprised the 3-min all-out exercise test (3 MT), with the initial 3 MT serving as familiarisation (Pettit et al. 2012). The 3 MT provides valid and reliable estimates of Vcrit (CV = 3.32–4.76%, ICC = 0.88–0.93) (de Aguiar et al. 2018). Both protocols were preceded by a standardised warm-up of 1600 m at a fixed velocity of 9 km h⁻¹ to preserve W’ and minimise priming effects (Bailey et al. 2009). This was followed by four standardised dynamic mobility exercises to prepare for the subsequent maximal effort. Participants were instructed to perform an all-out sprint effort in an anticlockwise direction on either of the two outermost lanes of a six lane 400-m athletics track. For both trials, participants began on the 300-m and 100-m start lines, respectively. Strong verbal encouragement was provided, although no information on elapsed or remaining time was given to discourage pacing. The 3 MT was terminated once 185-s had elapsed, to ensure that a complete 180-s period had been obtained. The mean velocity achieved during the final 30 s of the test was determined as Vcrit. Velocity data derived from GPS were modelled to determine mechanical work (J) and over-ground power (W) (refer to ‘Data analysis’ section). External power output associated with CP (W) was determined by the mean power output during the last 30 s, while W’ (kJ) was calculated as work performed (kJ) > CP. Criteria for re-test were applied as follows: (1) Vcrit achieved did not exceed 50% ∆ between VGET and vVO₂max (Pettit et al. 2012); and (2) between-trial coefficient of variation (CV) > 5% for Vcrit. Two participants re-tested due to violation of the second criteria.

30–15 intermittent fitness test

Visit 4 consisted of the 30–15 intermittent fitness test (30–15 IFT) (Buchheit, 2008). Testing took place on a synthetic athletics track. End-stage velocity (VIFT) validity and reliability has been well established with a typical error of 0.36 km h⁻¹ (Scott et al. 2015). Thus, an increase of one stage (0.5 km h⁻¹) was considered as meaningful. Participants were required to complete 30 s of running between a 40-m shuttle, interspersed by 15-s rest. The test began at 8 km h⁻¹ and progressed in 0.5-km h⁻¹ increments. Testing was terminated upon failing to sustain the required velocity or be within the required 3-m zone on three consecutive audio cues. The last fully completed stage was taken as VIFT (km h⁻¹), with Tlim measured in s.

Experimental trials

The experimental trials comprised five separate runs to exhaustion (Fig. 1) performed on a 400-m outdoor athletics track. For all trials, participants started on the 100-m start line and ran in an anticlockwise direction. Each trial was preceded by the standardised warm-up outlined above. The fifth visit comprised a constant work-rate trial in the severe domain (S_CWR) at vVO₂max. The final four trials (visits 6–9) comprised intermittent runs to exhaustion, consisting of 60-s work at vVO₂max + 50% Δ1 (where Δ1 = vVO₂max − Vcrit), interspersed with 30 s at a lower velocity determined by each of four different protocols, calculated as follows:

1. Light-domain recovery (S_L) at 40% vVO₂max.
2. Moderate-domain recovery (S_M) at 90% VGET.
3. Heavy-domain recovery (S_H) at VGET + 50% Δ2 (where Δ2 = Vcrit − VGET).
4. Severe-domain recovery (S_S) at vVO₂max − 50% Δ1 (where Δ1 = vVO₂max − Vcrit).

A whistle was blown at the start and end of each respective time interval. This corresponded to a change in velocity as signified by pre-recorded audio cues, which were projected through the portable amplifier. The trial was terminated upon volitional exhaustion or an inability to maintain the required velocity for three consecutive 10-m intervals. A handheld stopwatch was used to measure Tlim. Participants were not informed of elapsed time and work/recovery intensities, and no verbal encouragement was given to reduce confounding effects of motivation on Tlim (Andreacci et al. 2002).
Data analysis

Modelling over-ground power

For each protocol, raw velocity data sampled at 10 Hz were downloaded and exported to be processed in Microsoft Excel (Microsoft Corp., Redmond, USA). From this, estimations of work done were performed using an energetics model that has previously been applied to running in team sports (Furlan et al. 2015; Cummins et al. 2016; Gray et al. 2018). Briefly, by drawing upon principles of the work-energy theorem, this model assumes the runner as a multi-segment system of stature and mass, whereby metabolic energy demand (Eq. 1) is determined by the summation of total mechanical work ($W_{\text{tot}}$), partitioned into external work ($W_{\text{ext}}$) and internal work ($W_{\text{int}}$):

$$W_{\text{tot}} = W_{\text{ext}} + W_{\text{int}}.$$

These were then calculated as follows:

$$W_{\text{ext}} = W_{\text{hor}+} + W_{\text{hor}-} + W_{\text{vert}+} + W_{\text{vert}-} + W_{\text{air}},$$

(2a)

$$W_{\text{int}} = W_{\text{limbs}+},$$

(2b)

where $W_{\text{hor}+}$ and $W_{\text{hor}-}$ comprised work done (J kg$^{-1}$) when the centre of mass (COM) is, respectively, accelerated and decelerated in the horizontal plane. $W_{\text{vert}+}$ and $W_{\text{vert}-}$ comprised work done (J kg$^{-1}$) when the COM is, respectively, raised and lowered with each step in the vertical plane, $W_{\text{air}}$ comprised work done (J kg$^{-1}$) to overcome air resistance, and $W_{\text{limbs}+}$ comprised work done (J kg$^{-1}$) to swing the limbs with each step (Eqs. 2a and 2b). Resultant kinetic energy cost of the aforementioned variables, alongside duty factor (percentage of stride cycle spent in stance phase for a single limb) and stride frequency, was all assumed from GPS-derived velocity and acceleration. Therefore, starting with knowledge of forward running velocity, $W_{\text{tot}}$ was summed to ascertain total energy expenditure (J), from which over-ground power (W) was derived by dividing $W_{\text{tot}}$ from sampling duration (10-Hz). This modelling was applied to raw velocity data from the 3 MT, the 30–15 IFT and five runs to exhaustion to derive instantaneous power output and mechanical work. This permitted calculation of $T_{\text{LIM}}$ during the $S_{\text{CWR}}$ (constant speed) trial through rearrangement of the original Monod and Scherrer (1965) equation:

$$T_{\text{LIM}} = W'/ (P - CP),$$

(3)

where $T_{\text{LIM}}$ is time to exercise intolerance (s), $W'$ is finite work capacity (kJ), $P$ is external power output (W) during exercise, and CP is external power output associated with critical power (W). For intermittent trials, the difference between prescribed recovery power and CP ($D_{CP}$) was calculated from the designated 30-s recovery intensities. This was compared with the difference between actual $D_{CP}$, which was extrapolated from modelled over-ground power output.

Modelling $W'$ kinetics

The kinetics of $W'$ expenditure and reconstitution during intermittent protocols (visits 4, 6–8) were calculated at 0.1-s time intervals drawing upon the integral (Skiba et al. 2012) and differential (Skiba et al. 2015) equations for modelling of $W'_{\text{BAL}}$. The integral method ($W'_{\text{BALint}}$) was calculated as follows:

$$W'_{\text{BALint}} = W' - \int_{0}^{t} (W'_{\text{exp}}) (e^{-(t-u)/\tau_{\text{w}}}),$$

(4)

where $W'$ represents work done (kJ) > CP during the 3 MT, $W'_{\text{exp}}$ is the expenditure of $W' (t-u)$ is the time (s) between
segments that resulted in $W$ expenditure during the exercise bout, and $t_{w}$ is the time constant (s) for $W$ reconstitution. Thus, $W_{\text{BALdiff}}$ was the difference between available $W$ at the start of exercise and $W$ expenditure before time $t$. When external power output was below CP, $W$ was being reconstituted exponentially and calculated as follows:

$$\tau_{w} = 316e^{-0.01D_{CP}} + 316.$$  \hspace{1cm} (5)

The kinetics of $\tau_{w}$ vary curvilinearly as a function of $D_{CP}$ where numerical values are arbitrary time constants determined by nonlinear regression (Skiba et al. 2012). The 316 integral represents an asymptote beyond which a larger $D_{CP}$ does not facilitate further increases in $\tau_{w}$. The differential method ($W_{\text{BALdiff}}$) used for calculating $W$ expenditure was as follows:

$$W_{\text{BALdiff}}' = W_{0} - W_{\text{exp}}e^{-D_{CP}/W_{0}},$$  \hspace{1cm} (6)

where $W_{0}$ represents the available $W$ (kJ) at the start of the work segment; P and CP denote the participant’s external power output and external power output associated with critical power ($W$), respectively; $(t - u)$ is time (s) between segments that resulted in $W$ expenditure during which external power output exceeded CP. Reconstitution of $W$ using the differential method was calculated as follows:

$$W_{\text{BALdiff}}' = W_{0} - W_{\text{exp}}'e^{-D_{CP}/W_{0}},$$  \hspace{1cm} (7)

where $W_{0}$ represents the participant’s initial $W$ (kJ) as measured during the 3 MT, $W_{\text{exp}}'$ is $W$ expended as outlined in Eq. 6, and $D_{CP}$ is the difference between CP and external power output during the recovery segment.

### Statistical analyses

Statistical analyses were performed using SPSS version 24 (IBM SPSS Statistics Inc, Armonk, USA). Using Eq. 5, time constants for each participant during the $S_{L}$, $S_{M}$, and $S_{H}$ trials were varied iteratively until modelled $W_{\text{BAL}} = 0$ at exercise intolerance. The relationship between derived time constants ($\tau_{w}$) and $D_{CP}$ was analysed by nonlinear regression using GraphPad Prism 8 for macOS (GraphPad Software, San Diego, CA, USA). From this, an alternative exponential decay equation of the form $y = ae^{-b} + b$ was generated for $\tau_{w}$, thus obtaining a locomotor-specific $W_{\text{BAL}}$ model (OG-$W_{\text{BAL}}$) that was retrospectively applied to the 30–15 IFT. For the $S_{CWR}$ trial, differences between predicted and actual $T_{LIM}$ exponential analysis of variance (ANOVA) was performed, comparing actual $T_{LIM}$ and predicted $T_{LIM}$, as determined from the time (s) at which $W_{\text{BAL}} = 0$, respectively. Significant main effects were followed up with post hoc pairwise comparisons, using Bonferroni adjustments. Effect sizes were calculated using partial eta-squared ($\eta_p^2$) based on the following criteria: 0.02, small; 0.13, moderate; 0.26, large, or Cohen’s ($d$) for post-hoc pairwise comparisons: 0.2, small; 0.5, moderate; 0.8, large (Cohen 1988). Bivariate correlations ($r$) were performed between $\tau_{w}$ and $D_{CP}$ and all predictions of $T_{LIM}$ (actual, $W_{\text{BALint}}$, $W_{\text{BALdiff}}$). Using the following criteria: $\leq 0.1$, trivial; $> 0.1$ to 0.3, small; $> 0.3$ to 0.5, moderate; $> 0.5$ to 0.7, large; $> 0.7$ to 0.9, very large; and $> 0.9$ to 1.0, almost perfect (Hopkins et al. 2009). Statistical significance was accepted at $P < 0.05$ and data are reported as mean ± SD.

### Results

The participants’ $W$, CP, $VO_{2\text{max}}$, $\dot{V}O_{2\text{max}}$, and $V_{IFT}$ are presented in Table 1. During the $S_{CWR}$ trial, no significant differences were found between predicted and actual $T_{LIM}$ (514 ± 156 vs. 417 ± 76-s, $P = 0.094$, $d = 0.8$). Differences were found between prescribed and actual power outputs for the $S_{CWR}$ trial (511.9 ± 40.3 vs. 535.4 ± 49.8 W, $P < 0.001$, $d = 0.5$), as well as between prescribed and actual $D_{CP}$ in the $S_{M}$ (47.9 ± 13.7 vs. 42.1 ± 11.3 W, $P = 0.031$, $d = 0.5$) and $S_{H}$ (17.6 ± 8.8 vs. 34.9 ± 5.2 W, $P < 0.001$, $d = 2.4$) trials, but not for the $S_{L}$ (167.6 ± 73.8 vs. 150.4 ± 28.7 W, $P = 0.441$, $d = 0.3$). The main experimental trials from participant 8 are presented in Fig. 2.

Nonlinear regression analysis conducted on $\tau_{w}$ as a function of $D_{CP}$ (Fig. 3) yielded a moderate relationship with $S_{L}$, $S_{M}$, and $S_{H}$ trials ($r^2 = 0.52$). Six outliers were removed from the analysis due to wind speeds exceeding 28.8 km h$^{-1}$ on the day of trials, which led to unfeasible $D_{CP}$ and $\tau_{w}$ values. Using an exponential one-phase decay method, the data were best fit by the following equation:

$$\tau_{w} = 372e^{-0.02D_{CP}} + 102,$$  \hspace{1cm} (8)

where, 372 is an arbitrary value representing the difference between $\tau_{w}$ when $D_{CP}$ = 0 and the derived plateau, 0.02 is the derived rate constant expressed as a reciprocal of $D_{CP}$ and 102 represents the asymptote time constant, beyond which a larger $D_{CP}$ facilitates no further increases in $\tau_{w}$. The $\tau_{w}$ for the $S_{L}$, $S_{M}$ and $S_{H}$ were 119 ± 32 s, 190 ± 45 s, and 336 ± 77 s, respectively (Table 2). In addition, $\tau_{w}$ was inversely correlated with $D_{CP}$ in the $S_{L}$, $S_{M}$, and $S_{H}$ trials ($r = -0.66, 95\% \text{CI: } -0.85–0.32, P = 0.001$).

Equation 8 was retrospectively applied to the 30–15 IFT and compared to $W_{\text{BALint}}$ and $W_{\text{BALdiff}}$. Results for
$T_{\text{LIM}}$ as predicted by the three $W'_{\text{BAL}}$ models are presented in Table 3. Repeated-measures ANOVA revealed differences between actual and predicted $T_{\text{LIM}}$ across groups ($F_{(1,397,11,175)} = 25.248$, $P < 0.001$, $\eta^2_p = 0.759$). Post hoc pairwise comparisons revealed a significant difference between actual $T_{\text{LIM}}$ (968 ± 117-s) and $T_{\text{LIM}}$ predicted by $W'_{\text{BALint}}$ (848 ± 91-s, $P = 0.001$, $d = 1.1$). There were no differences between actual $T_{\text{LIM}}$ and those predicted by $W'_{\text{BALdiff}}$ (938 ± 84-s, $P > 0.100$, $d = 0.3$) and the OG-$W'_{\text{BAL}}$ (929 ± 94-s, $P > 0.100$, $d = 0.4$), with an almost perfect correlation between $T_{\text{LIM}}$ predicted by $W'_{\text{BALdiff}}$ and OG-$W'_{\text{BAL}}$ during the 30–15 IFT ($r = 0.98$, 95% CI: 0.91–1.00, $P < 0.0001$). There were also very strong correlations between actual and predicted $T_{\text{LIM}}$ from the OG-$W'_{\text{BAL}}$ ($r = 0.88$, 95% CI: 0.53–0.98, $P = 0.002$) and $W'_{\text{BALdiff}}$ ($r = 0.88$, 95% CI 0.50–0.97, $P = 0.002$) methods. The mean difference between actual $T_{\text{LIM}}$ and that predicted by OG-$W'_{\text{BAL}}$ and $W'_{\text{BALdiff}}$ was 39.4 and 29.6 s, respectively. Modelled $W'_{\text{BAL}}$ on the 30–15 IFT for two representative participants is presented in Fig. 4.

### Discussion

This study mathematically characterised the kinetics of $W'_{\text{BAL}}$ during intermittent over-ground running through the development of a locomotor-specific model (OG-$W'_{\text{BAL}}$). This offered accurate predictions of $T_{\text{LIM}}$ (mean difference 39.4 s) during severe-intensity intermittent running to exhaustion (Table 3). An exponential relationship was also established between $v_{\text{VO2max}}$ and $D_{\text{CP}}$ across the $S_M$, $S_H$, and $S_L$ exercise intensity domains (Fig. 3), as reported during indoor cycling (Skiba et al. 2012). An additional finding of this study was the strong relationship between the OG-$W'_{\text{BAL}}$ and $W'_{\text{BALdiff}}$ for predicting $T_{\text{LIM}}$ in the 30–15 IFT ($r = 0.98$).

Although the new OG-$W'_{\text{BAL}}$ modelled $W'$ kinetics in a similar way to the current differential equation, with no differences to actual intermittent $T_{\text{LIM}}$, the $T_{\text{LIM}}$ predictions were not perfect and, therefore, require further discussion. The discrepancies between the prescribed and actual external power outputs from both the $S_{\text{CWR}}$ and intermittent trials indicate problems with sustaining the appropriate steady-state speeds during over-ground running (Fig. 2). This possibility was considered prior to the study, and we accounted for potential pacing irregularities by cuing participants according to pre-programmed audio signals and a priori familiarisation to the protocol. Despite these measures, we found differences between predicted and actual $T_{\text{LIM}}$ during the $S_{\text{CWR}}$ trial (514 ± 156 vs. 417 ± 76 s, respectively), which we attribute to 23.5 W discrepancies in external power output (prescribed 511.9 ± 40.3 vs. actual 535.4 ± 49.8 W). Predictions of constant-intensity exercise were based on a rearrangement of the linear equation of Monod and Scherrer (1965). Therefore, reliable representations of steady-state external power output were necessary during the $S_{\text{CWR}}$. Our results were most likely attributable to the observed differences in external power output during $S_{\text{CWR}}$. In support of this, faster constant running or cycling at $v_{\text{VO2max}}$ (i.e., no pacing control) is associated with reduced $T_{\text{LIM}}$, attributable to more rapid $W'$ expenditure, despite a fixed work capacity (Billat et al. 1994; Chidnok et al. 2013a). Similar discrepancies were found during the intermittent trials, where differences between prescribed and actual $D_{\text{CP}}$ were found during the $S_M$ (48 ± 14 vs. 42 ± 11 W, respectively) and $S_H$ trials (18 ± 9 vs. 35 ± 5 W, respectively). Interestingly, there was a trend towards a reduced $D_{\text{CP}}$ difference in $S_M$ trials, whilst there was an increased $D_{\text{CP}}$ difference in $S_H$ trials, producing a mean difference in actual $D_{\text{CP}}$ between the $S_M$ and $S_H$ of 7.2 W. Whilst this difference was seemingly able to demarcate the moderate- and heavy-intensity domains,
as evidenced by \( t_w \) of 190 ± 45 vs. 336 ± 77 s for \( S_M \) and \( S_H \) respectively, it meant that the \( D_{CP} \) for these intensities varied from that intended and might have affected the OG-WBAL modelling.

Variability in pacing is inevitable during the trials and has logical implications for modelling of over-ground power, as changes in horizontal velocity, and thus kinetic energy (accelerating or decelerating), on level surfaces partially determines horizontal work done (\( W_{\text{hor}} \)). Since \( W_{\text{hor}} \) contributes to total external work done (\( W_{\text{ext}} \)) in the current over-ground power model (Gray et al. 2018), any velocity changes will lead to fluctuations in external power output. Whilst some variability in pacing is expected, modelling of over-ground power is sensitive to velocity changes. This is of greater importance during the current exercise model, because missing a cone during the trial would most likely prompt the participant to sharply increase their speed to ensure that the subsequent cone was reached in time. The accumulation of these minor velocity changes would increase the work done and, therefore, external power output during the trials above that prescribed. Indeed, using this reasoning, the closer
matching of prescribed and actual $D_{CP}$ during the $S_L$ and $S_M$ trials compared to the larger differences in the more intense $S_H$ trials is logical, since performing the $S_H$ trial requires greater accelerations and decelerations between work and recovery periods, respectively, with any pacing errors necessitating a rapid and more intense compensation in velocity. Collectively, our results provide evidence that the energetics model can be used to predict $T_{LIM}$ during constant or intermittent work-rate trials, but more

![Fig. 3](image)

**Fig. 3** The time constant for $W'$ reconstitution ($\tau_{W'}$) plotted as a function of the difference between recovery power output and critical power ($D_{CP}$). Individual reconstitution times are represented by a common symbol, where severe–light ($S_L$) = circles, severe–moderate ($S_M$) = triangles, and severe–heavy ($S_H$) = diamonds.

### Table 2

Calculations for the time constant of $W'$ reconstitution ($\tau_{W'}$) across all intermittent trials for each participant

| Participant | $S_L$ | $S_M$ | $S_H$ |
|-------------|------|------|------|
| 1           | 94   | –    | 350  |
| 2           | 104  | 182  | –    |
| 3           | 160  | 201  | 213  |
| 4           | 174  | –    | 412  |
| 5           | 135  | –    | –    |
| 6           | 86   | 137  | –    |
| 7           | 128  | 270  | 370  |
| 8           | 89   | 182  | 275  |
| 9           | 98   | 166  | 394  |
| Mean        | 119  | 190  | 336  |
| SD          | 32   | 45   | 77   |

Dashes denote six outliers that were removed from the data set and were, therefore, not included in the final analysis. $S_L$, severe–light domain intermittent trial; $S_M$, severe–moderate domain intermittent trial; $S_H$, severe–heavy domain intermittent trial.

### Table 3

Calculations for the time limit of tolerance ($T_{LIM}$) on the 30–15 intermittent fitness test (30–15 IFT) as predicted by modelling of $W'_{BAL}$ according to the integral ($W'_{BAL,int}$), differential ($W'_{BAL,diff}$), and locomotor-specific (OG-$W'_{BAL}$) methods

| Participant | Actual | $W'_{BAL,int}$ | $W'_{BAL,diff}$ | OG-$W'_{BAL}$ |
|-------------|--------|----------------|----------------|--------------|
| 1           | 1104   | 956            | 1068           | 1054         |
| 2           | 1024   | 863            | 925            | 926          |
| 3           | 839    | 789            | 847            | 852          |
| 4           | 784    | 707            | 841            | 813          |
| 5           | 979    | 816            | 916            | 877          |
| 6           | 928    | 865            | 940            | 924          |
| 7           | 929    | 866            | 966            | 953          |
| 8           | 1155   | 999            | 1067           | 1095         |
| 9           | 970    | 770            | 876            | 863          |
| Mean        | 968    | 848            | 938            | 929          |
| SD          | 117    | 91             | 84             | 94           |

**Fig. 4** $W'_{BAL}$ responses displaying the integral ($W'_{BAL,int}$), differential ($W'_{BAL,diff}$), and locomotor-specific (OG-$W'_{BAL}$) methods for participant 1 (a) and participant 6 (b) during the 30–15 intermittent fitness test. Note how in b, OG-$W'_{BAL}$ reaches full depletion within 924 s (4 s from $T_{LIM}$) vs. 940 s and 865 s for $W'_{BAL,diff}$ and $W'_{BAL,int}$, respectively. For all participants, OG-$W'_{BAL}$ displayed very similar kinetics to $W'_{BAL,diff}$ as shown in a and b.
thorough ways of pacing athletes during the modelling process might enhance its precision.

The observed curvilinear relationship between $\tau_{W^*}$ and $D_{CP}$ is consistent with findings from Skiba et al. (2012) during a similar intermittent cycling trial. Interestingly, the $\tau_{W^*}$ reported were $377 \pm 29$-s for recovery at $20$ W, $452 \pm 81$-s for moderate-intensity recovery, and $578 \pm 105$-s for heavy-intensity recovery. These $\tau_{W^*}$ were considerably higher than those reported in the current study (Table 2). It is plausible that adjustments to systemic oxygen transport and muscle metabolism were apparent in response to the change in exercise modality (Caputo et al. 2003). Indeed, development of the $\dot{V}O_2$ slow component is strongly related to type II muscle fibre recruitment, alongside increases in the rate of blood lactate and intramuscular metabolite accumulation, alongside PCR depletion (Poole et al. 2016). Jones and McConnell (1999) reported a significant difference in the $\dot{V}O_2$ slow component between cycling ($290 \pm 102$ mL·min$^{-1}$) and treadmill running ($200 \pm 45$ mL·min$^{-1}$) during heavy-intensity exercise, which was later corroborated by Carter et al. (2000).

The development of a smaller slow component has been historically linked to the different muscle contraction regimen of running locomotion compared to cycling (Jones and McConnell 1999; Carter et al. 2000; Hill et al. 2003), which involves a more pronounced eccentric component, requiring storage and utilisation of elastic energy (Komi 2000). This is in stark contrast to the concentrically biased actions of pedalling (Carter et al. 2000), where it is assumed that greater recruitment of type II motor units occurs in response to the higher intramuscular pressures and intermittent occlusion of blood flow, leading to greater metabolic perturbation (Caputo et al. 2003). Similar reasoning and findings were reported in inclined compared to flat locomotion, where the stretch shortening cycle activity is reduced in preference for prolonged concentric muscle actions during uphill running (Pringle et al. 2002). The smaller slow component anticipated during running compared to cycling, therefore, provides one possible explanation for the smaller $\tau_{W^*}$ in the current study relative to others (Skiba et al. 2012).

The reported discrepancies between actual $T_{LIM}$ during the 30–15 IFT ($968 \pm 118$-s) and predicted $T_{LIM}$ from $W_{BAL,lim}$ ($848 \pm 91$-s) can be attributed to the mode-specific averaged time constants for cycling used in the original model. These outcomes were anticipated by Skiba et al. (2015), prompting the development of the $W_{BAL,diff}$. The $W_{BAL,diff}$ uses a mathematical framework that permits scaling of recovery kinetics to the power output of the exercise modality, such that the fitting of specific time constants can be negated. The close comparisons ($T_{LIM}$ differences ~ 9-s) and temporal matching of the $W_{BAL,diff}$ and the OG-$W_{BAL}$ ($r=0.98$) are, therefore, remarkable. It was unclear how strong this relationship would be, since the $W_{BAL,diff}$ was originally modelled on single-leg extensor exercise (Skiba et al. 2015), compared to the mode-specific integral model reported herein (OG-$W_{BAL}$).

However, it was hypothesized that the new OG-$W_{BAL}$ would predict $T_{LIM}$ accurately during the 30–15 IFT, since its estimations were based on the same group of participants. Indeed, the strong relationships between these two models are most likely attributed to this, which indicates the need to apply the new OG-model to other participants during intermittent running tasks of varying work-recovery composition. While the close agreement between the OG-$W_{BAL}$ and $W_{BAL,diff}$ estimation of $T_{LIM}$ indicates that both methods are capable of accurately estimating intermittent running energetics, the behaviour of the differential model during whole-body exercise of substantially higher absolute power outputs, yet lower $D_{CP}$ values compared to cycling or single-limb movements (Skiba et al. 2012, 2015), supports its wider application to other modes of human locomotion.

**Limitations**

The validity of modelled over-ground power output and $W_{BAL}$ kinetics in this study rely upon the assumption that derived locomotive mechanical work is performed on a uniform flat surface. The extension of the OG-$W_{BAL}$ and $W_{BAL,diff}$ models to non-uniform gradients is, therefore, currently restricted. Moreover, robustness of the OG-$W_{BAL}$ model will be predicated upon its extended application to more diverse populations than used in the current study. However, the temporal behaviour and close proximity of OG-$W_{BAL}$ to $W_{BAL,diff}$ provides preliminary evidence of its validity in this group of participants.

**Practical applications**

The power–duration relationship and $W_{BAL}$ modelling in over-ground running have practical applications to intermittent sports, where exercise is performed below (i.e., recovery) and above the severe-intensity domain. These include sports such as football, rugby, and field hockey, where over-ground speed is often used as a proxy of exercise intensity. However, quantification of work done at low velocity is problematic using traditional kinematic approaches, but can be overcome by the application of mechanical models (Gray et al. 2018). Integration of such models with knowledge of the athlete’s CP and $W^*$ would permit team-sport practitioners to profile athletes based on individual characterisation of the power–duration relationship and monitor these changes throughout the season. Furthermore, using the equations developed herein or previously (OG-$W_{BAL}$ or $W_{BAL,diff}$), the internal workload of a player could be quantified non-invasively based on a universal energetic metric of ‘work done’ (J) in physiologically relevant exercise domains. Analysis of $W_{BAL}$ using one of the above models may also provide insights into transient
fatigue and variation in work rate during intermittent exercise performance. Thus, this approach could help in determining the intensity of training sessions and acutely predict exercise intolerance during repeated high-intensity running. Before this can be achieved, further work is required to validate the current model via its application to other forms of intermittent over-ground exercise and larger or more diverse samples.

Conclusion

Our findings demonstrate, for the first time, that $W_{\text{BAL}}$ can be modelled during intermittent over-ground running. A locomotor-specific integral equation (OG-$W_{\text{BAL}}$) was able to accurately track the expenditure and reconstitution of $W'$, such that its depletion approximated participant exhaustion during severe-intensity intermittent running. The OG-$W_{\text{BAL}}$ and $W_{\text{BAL} \text{diff}}$ methods performed similarly in modelling $W'$ kinetics; therefore, either equation would provide equally robust estimations. However, the use of $W_{\text{BAL} \text{int}}$ for over-ground running is not recommended. These findings provide scope for studies and practitioners to implement $W_{\text{BAL}}$ modelling to more objectively quantify the internal cost of field-based intermittent running activities.

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Author contributions

CV and MW conceived and designed research. CV conducted the experiment. All authors’ analysed data, wrote, read, and approved the manuscript.

Compliance with ethical standards

Conflict of interest

The authors declare that they have no conflict of interest.

Ethical approval

All procedures were performed in accordance with the ethical standards of the institutional and/or national research committee (St Mary’s University Ethics Committee, ref: SMEC_2018-19_056).

Informed consent

Written informed consent was obtained from all individual participants included in the study.

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