



## Abstract

**Objectives:** To describe the perceived load, fatigue and recovery profiles during congested and non-congested schedules in international football tournaments. **Design:** Retrospective single-cohort observational study **Methods:** Internal load (session-rating of perceived exertion [s-RPE]) and perceived ratings of fatigue, muscle soreness, psychological status, sleep quality, and sleep duration were recorded daily from 37 national team footballers during the competition phase of 3 international tournaments. ANOVA and Effect Size (ES) analyses compared individualised internal load and perceived response profiles between congested and non-congested acute 2-match schedules. Conditions included Acute Congestion ( $\leq 4$  days between two matches), Non-Congestion ( $> 4$  days between two matches), Single-Match, and No-Match. **Results:** Significantly higher s-RPE match loads ( $p < 0.001$ ) within the single- and multi-match conditions resulted in significantly worsened ( $p < 0.05$ ) subjective ratings of perceived fatigue, muscle soreness and sleep duration in the 24-48h post-match. Internal load profiles were not different between the Acute-Congestion or Non-congestion conditions ( $p > 0.05$ ); though Acute-Congestion had significantly worsened pre-match subjective ratings compared to Non-Congestion on both MD1 ( $p = 0.040$ ; ES = 0.94) and MD2 ( $p = 0.033$ ; ES = 0.94). However, between-match differences in Acute-Congestion showed no further impairments in perceived response between the first and second matches ( $p > 0.05$ ). **Conclusion:** During international tournaments, internal load and perceived fatigue/recovery profiles are largely determined by their exposure (or lack thereof) to match-play. Periods of acute match congestion impaired players pre-match perceived status when compared to non-congested microcycles. However, acute match congestion does not appear to exacerbate players post-match fatigue/recovery response within the context of international football tournaments.

**Keywords:** training load; fatigue; recovery; monitoring; national team; international football

## 27    **Introduction**

28    Major international football tournaments consist of demanding fixtures, with teams required  
29    to play 3 matches within 7-11 days, and those in knockout phases playing up to 7 matches  
30    over 28-31 days.<sup>1</sup> To optimise tournament performance, balancing the stress of  
31    training/competition with sufficient recovery is a major objective for support staff.<sup>2</sup> Such  
32    views are articulated in a survey at the 2014 World Cup, whereby national team practitioners  
33    ranked reduced recovery periods, accumulated fatigue, and congested match schedules among  
34    the most important risk factors for non-contact injury.<sup>3</sup> Similarly, national team practitioners  
35    also ranked monitoring tools that quantify match exposure and subjective markers of fatigue  
36    and recovery status as commonly used.<sup>3</sup> However, whilst player monitoring practices are  
37    ubiquitous within professional football,<sup>3,4</sup> evidence of their usefulness emanates  
38    predominately from club-based contexts,<sup>5-7</sup> with limited research available within national  
39    team environments.

40  
41            A common trend in the monitoring practices of elite football teams is the daily use  
42    of athlete self-report measures to monitor the psychobiological state or “wellness” of  
43    players.<sup>3,4</sup> Although the validity of such scales remains in discussion<sup>8,9</sup>, the applicability of  
44    monitoring athlete perceived stress and fatigue in professional football has some support, with  
45    studies demonstrating their responsiveness to changes in acute training and match loads.<sup>2,10,11</sup>  
46    For instance, within an elite professional team, between-day changes in subjective measures  
47    of perceived fatigue, muscle soreness and sleep quality were shown to closely reflect acute  
48    fluctuations in training/match load across ‘standard’ in-season weeks (1-match per week).<sup>10</sup>  
49    Specifically, significant reductions in total wellness (35-40%) exist the day post-match, with  
50    subsequent improvements (17-26% increase) by 48h and a plateau thereafter due to the  
51    renewed presence of training load.<sup>10</sup> However, despite the apparent responsiveness of these  
52    subjective measures to the acute presence of training/match load, the association between load  
53    and wellness in football remains equivocal, with contrasting findings depending on the load

variable measured (i.e. internal or external load) and how load is calculated (i.e. previous day load, accumulation of load over multiple-days/week).<sup>2,11,12</sup>

While the aforementioned research examined the perceived fatigue and stress responses of professional footballers during single-match week microcycles, the competition schedules of most successful football teams includes multi-match weeks, as exists during international tournaments.<sup>13,14</sup> Within congested match schedules the time for recovery is reduced, accentuating the need to optimise and monitor recovery process. However, limited research exists detailing the impact of match congestion on player recovery profiles in international competition. Recent studies within club football have shown ratings of perceived fatigue to be significantly worsened in weeks with two-competitive matches compared to one.<sup>15,16</sup> While, gradual decreases in match day total wellness have been observed over a prolonged period of fixture congestion (11 matches in 36 days), indicating that subjective status may be sensitive to chronic accumulations of load.<sup>17</sup> Thus far, no study has examined the acute load and recovery profiles of international footballers during congested match-to-match microcycles. Accordingly, this study examined the effect of match load on self-reported fatigue and recovery profiles during congested and non-congested microcycles within international tournaments. It is hypothesised that match exposure during a congested microcycle will negatively impact perceived recovery profiles when compared to non-congested microcycles.

## Methods

Data was collected from 37 professional male football players ( $26.4 \pm 4.1$ y) selected to compete for the Australian National Football Team at three Fédération Internationale de Football Association (FIFA) sanctioned international football tournaments: 2015 Asian Cup (AC<sub>2015</sub>), 2017 Confederations Cup (CC<sub>2017</sub>), and 2018 World Cup (WC<sub>2018</sub>). All outfield players from the original 23-man squad for each tournament were eligible for inclusion, with goalkeepers excluded due to variations in their training methods and match activity. In total, participating players consisted of 8 central defenders, 6 wide defenders, 12 central midfielders, 6 wide midfielders and 5 strikers. Provisional approval for the study was obtained from the National Federation involved, with individual player data previously collected as a condition of national team duty.<sup>18</sup> Data collection procedures were approved by the institutional Human Research Ethics Committee. Retrospective sharing and conditional usage of the data was undertaken in accordance with a strict data confidentiality agreement, and all data were anonymised before analysis to ensure player confidentiality.

The present observational study followed a retrospective single-cohort study design, in which players' internal load (training and matches) and perceived fatigue and recovery data were routinely collected from the same national team competing at three international tournaments. Data was assessed from only the competition phases of each tournament, comprising 23-days for the AC<sub>2015</sub> (winners), 7-days for the CC<sub>2017</sub> (3<sup>rd</sup> in group stages), and 12-days for the WC<sub>2018</sub> (4<sup>th</sup> in group stages). In total 12 matches were played across the 3 tournaments, in which the observed national team registered 5 wins, 3 draws and 4 losses. Individual player tournament data was then categorised into congested or non-congested 2-match microcycles (i.e. consecutive matches), according to the following classification criteria: 1) Acute-Congestion – participation in 2 successive matches separated by a time interval totalling < 4 days (96h)<sup>19</sup>; 2) Non-Congestion – participation in 2 successive matches separated by a time interval totalling > 4 days (>96h); 3) Single-Match week – 1 match

played followed by no-participation in the subsequent match; 4) No-Match week – no participation in successive matches. Match participation was defined as having played  $\geq 70$  minutes in a single match.<sup>14,16</sup> For instances where microcycles overlapped (i.e. two Acute-Congestion microcycles from 3 consecutive matches) data from the middle match was included within both microcycles. Overall, a combined total of 70 match-to-match microcycles were included for analysis, consisting of 18 Acute-Congestion, 19 Non-Congestion, 10 Single-Match, and 23 No-Match microcycles.

Internal player load for all training and match sessions (including on-field, gym, recovery and rehab training sessions) were monitored throughout the assessment period using the session-rating of perceived exertion (s-RPE) method.<sup>20</sup> Data were manually entered by the players into a tablet application (SMARTABASE, Fusion Sport, Brisbane, Australia), in which the same reporting scale (Modified CR-10 Borg scale)<sup>21</sup> and recording protocol were used across all international tournaments. All players had familiarity with the scale, having used such method for at least 1-year prior. Session load (s-RPE, arbitrary units, AU) was then estimated for each player by multiplying their total training or match session duration (min) by their reported RPE.<sup>20</sup>

Player's perceived ratings were monitored daily each morning (09:00-10:00) prior to training or competition throughout the assessment period. A customised questionnaire measured player's perceived ratings to assess their fatigue and recovery responses to training and competition.<sup>22-24</sup> It comprised 5 items relating to perceived fatigue, general muscle soreness, psychological status, sleep quality, and sleep duration. Each question was scored on a seven-point Likert scale ranging from 1–7, with point increments of 1, and scores of 1 representing “very, very good” and 7 representing “very, very poor”, respectively. The five items for a given day were summated to provide a total perceived response score for each player, whereby a higher number represents a worse perceived state.

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129           Descriptive data for internal load and perceived fatigue/recovery response variables  
130 are presented as means  $\pm$  standard deviation (mean  $\pm$  SD). Due to differences in the time  
131 intervals between matches (i.e. 3, 4, or 5-days), individual load and perceived response data  
132 were aligned within each 2-match microcycle according to the number of days post-match  
133 (i.e. matchday 1 [MD1], matchday 1+1 [MD1+1], matchday 1+2 [MD1+2], matchday 1+3  
134 [MD1+3], and repeated for matchday 2 [MD2]). Subsequent analysis of the training load  
135 profile was performed using data up 2-days post-match, while player perceived ratings were  
136 analysed using data up to 3-days post-match to assess the full extent of the player responses as  
137 a proxy for recovery.

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139           Mixed-design analysis of variances (ANOVA) determined the effect of different  
140 acute match congestion microcycles on player's load and perceived fatigue and recovery  
141 profiles, with time as the within-subjects factor and microcycle condition as the between-  
142 subjects factor. Subsequent follow-up tests were conducted using multiple one-way ANOVAs  
143 to determine where these difference/s occurred, with significance set at the  $p \leq 0.05$  level. The  
144 magnitude of these differences was reported as Cohen's  $d$  effect sizes (ES) with [90%  
145 confidence intervals]. The criteria used to interpret the magnitude of the ES were as follows:  
146 0.0–0.2 = trivial, 0.21–0.6 = small, 0.61–1.2 = moderate, 1.21–2.0 = large, 2.01–4.0 = very  
147 large, >4.0 = nearly perfect.<sup>25</sup> Additionally, Spearman's rank correlation coefficients ( $r_s$ ) were  
148 also calculated to determine the association between internal match load and self-reported  
149 variables the day immediately post-match (MD+1). Initial correlations were performed using  
150 data from all 2-match microcycles, followed by a secondary analysis in which match loads of  
151 zero (i.e. No Match condition) were excluded. The magnitude of the correlations was  
152 classified as follows: trivial ( $r_s \leq 0.1$ ), small ( $0.1 < r_s \leq 0.3$ ), moderate ( $0.3 < r_s \leq 0.5$ ), large ( $0.5 <$   
153  $r_s \leq 0.7$ ), very large ( $0.7 < r_s \leq 0.9$ ), almost perfect ( $r_s > 0.9$ ).<sup>25</sup> All statistical analyses were  
154 performed using the software package SPSS (version 24.0, Chicago, IL, USA).

## 155     **Results**

156     Concerning internal load, both main effects of microcycle condition ( $F_{[3, 66]} = 61.25, p < 0.001$ )  
157     and time ( $F_{[3, 177]} = 439.77, p < 0.001$ ) were significant across the 2-match microcycle periods.  
158     The interaction between microcycle condition and time was also significant ( $F_{[8, 177]} = 197.01,$   
159      $p < 0.001$ ); with post-hoc comparison tests revealing significantly higher ( $p < 0.001$ ; Figure 1a)  
160     match loads for Acute-Congestion (ES = 12.69[12.07,13.31]), Non-Congestion (ES =  
161     12.91[12.32, 13.50]), and Single-Match (ES = 12.36[11.29,13.44]) on MD1, and for Acute-  
162     Congestion (ES = 9.98[9.47,10.48]; ES = 12.22[11.60,12.84]) and Non-Congestion (ES =  
163     12.07[11.58,12.55]; ES = 14.69[14.09,15.28]) on MD2. Post-matchday, the No-Match  
164     condition reported significantly ( $p < 0.001$ ) higher training loads on MD1+1 compared to all  
165     other conditions (ES<sub>range</sub> = 3.02–3.11); while on MD2+1, training loads were significantly  
166     higher ( $p < 0.001$ ) for Single-Match (ES = -3.30[-4.29,-2.32]; ES = -3.14[-4.12,-2.16]) and  
167     No-Match (ES = -3.39[-3.87,-2.92]; ES = -3.30[-3.79,-2.82]) conditions compared to  
168     Acute-Congestion and Non-Congestion. On MD1+2, Non-Congestion reported significantly  
169     higher loads compared to Acute-Congestion ( $p = 0.063$ ; ES = -0.86[-1.41, -0.30]) and  
170     Single-Match ( $p = 0.009$ ; ES = 1.04[0.54,1.53]), while the No-Match group also had  
171     significantly higher ( $p = 0.011$ ) training loads compared to Single-Match (ES = -0.87[-1.31,  
172     -0.44]). Lastly on MD2+2, Acute Congestion reported significantly higher ( $p = 0.015$ )  
173     training loads compared to Non-Congestion (ES = 1.16[0.57,1.75]).

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175             Concerning total perceived responses, both main effects of microcycle condition  
176     ( $F_{[3, 66]} = 5.42, p = 0.002$ ) and time ( $F_{[4, 254]} = 57.56, p < 0.001$ ) were significant across the 2-  
177     match microcycle periods. The interaction between microcycle condition and time was also  
178     significant ( $F_{[12, 254]} = 4.77, p < 0.001$ ); with post-hoc comparison tests revealing significantly  
179     higher (worsened) total perceived responses for Acute-Congestion compared to Non-  
180     Congestion on MD1 ( $p = 0.040$ ; ES = 0.94[0.37,1.50]) and MD2 ( $p = 0.048$ ; ES =  
181     0.94[0.39,1.49]). Post-matchday, total response values were significantly higher for Acute-



Congestion ( $p < 0.001$ ; ES = 1.81[1.31,2.31]), Non-Congestion ( $p = 0.009$ ; ES = 1.04[0.52,1.56]), and Single-Match ( $p = 0.005$ ; ES = 1.21[0.65,1.77]) on MD1+1; while on MD2+1, total response values were significantly higher for Acute-Congestion and Non-Congestion compared to both Single-Match ( $p = 0.003$ ; ES = 2.20[1.36,3.05];  $p = 0.026$ ; ES = 1.51[0.70,2.32]) and No-Match ( $p < 0.001$ ; ES = 2.10[1.61,2.59];  $p < 0.001$ ; ES = 1.56[1.06,2.05]) conditions. Thereafter, total perceived response ratings remained significantly ( $p = 0.048$ ) higher for Acute-Congestion (ES = 0.81[0.31, 1.32]) compared to No-Match on MD1+2, while no significant differences ( $p > 0.05$ ) were evident on MD1+3. Lastly, on MD2+2, total perceived response remained significantly higher for Acute-Congestion ( $p = 0.005$ ; ES = 1.86[1.08,2.65]) and Non-Congestion ( $p = 0.017$ ; ES = 1.37[0.67,2.08]) compared to Single-Match, while no significant differences ( $p > 0.05$ ) between microcycle conditions were evident on MD2+3.

In relation to the perceived response sub-scales, significant interaction effects for perceived fatigue, muscle soreness, sleep duration and sleep quality were evident ( $p < 0.05$ ; Table 1), whereby Acute-Congestion and Non-Congestion showed significantly higher post-match ratings of perceived fatigue ( $p \leq 0.032$ ; ES<sub>range</sub> = 0.98–2.78) and muscle soreness ( $p \leq 0.024$ ; ES<sub>range</sub> = 1.00–3.33) up to 48h post-match. These significantly higher fatigue ratings remained elevated up to 72h post-match, although higher muscle soreness ratings were only evident at 72h post-MD2. In addition, Acute-Congestion reported significantly higher sleep duration and sleep quality on both MD1+1 and MD2+1 compared to No-Match.

Significant positive correlations ( $p < 0.001$ ; Figure 2) existed between s-RPE match load and the following post-match items: total perceived response ( $r_s = 0.571$ ), fatigue ( $r_s = 0.673$ ), muscle soreness ( $r_s = 0.675$ ), sleep quality ( $r_s = 0.364$ ), and sleep duration ( $r_s = 0.303$ ). However, when data points with zero match load were excluded (i.e. No-Match microcycles) and the correlation analysis repeated, only a small negative correlation ( $r_s = -$

209 0.255;  $p < 0.05$ ) was found between s-RPE match load and sleep duration (i.e. higher load and  
210 reduced sleep duration), while no significant correlations were identified for the other post-  
211 match perceived response items ( $p > 0.05$ ).

## Discussion

This study examined the perceived load, fatigue and recovery profiles of international footballers during congested and non-congested tournament schedules. Variations in load and player perceived status profiles were largely determined by exposure to match-play, in turn influencing acute post-match responses. As expected, high internal match loads preceded reduced next-day training loads in single and multi-match conditions. The resultant post-match responses showed worsened ratings of fatigue, muscle soreness, and sleep quality/duration within these 2-match microcycles. Furthermore, pre-match ratings for Acute-Congestion were significantly worse compared to Non-Congestion on both match days. However, within the Acute Congestion conditions no significant differences in perceived status were evident between respective match days. Within the context of international football tournaments, acute match congestion (i.e. 2 matches within 4 days) does not appear to exacerbate post-match perceptual fatigue and recovery responses.

Despite an abundance of club-based data, limited match load and recovery data exists from national team contexts. In the present study, exposure (or lack thereof) to tournament match-play largely determined the change in load profiles between match congestion microcycles. Likely at the instigation of coaching staff, players exposed to >70min playing time resulted in lower training loads over the ensuing days compared to non-playing condition. Although intuitive, this finding reflects a common post-match squad load management strategy, whereby recovery is emphasised for starters/substitutes and maintenance of physical loading for non-players.<sup>26,27</sup> Regardless of the existence or absence of match congestion, similar internal load profiles existed between matches, which concurs with findings from club-based contexts.<sup>16</sup> In further support, external locomotion-based variables have also reported to be not significantly different during congested schedules in both professional<sup>19,28</sup> and international football.<sup>1</sup> Thus, while the absence of external load measures is a potential limitation, the similarities in internal load responses between the

conditions suggest an appropriate reference point from which to compare the post-match perceptual recovery profiles.

Differences in perceived fatigue and recovery profiles between the match congestion conditions closely reflected exposure to tournament match-play, hence questioning whether exposure or load are of importance. For instance, similar impairments in total perceived response values were observed up to 48h post-match in both single and multiple matches. Previous research reporting perceptual wellness responses within ‘standard’ 1 match microcycles<sup>2,10</sup>, as well as, both single and multi-match weeks,<sup>16</sup> reported comparable post-match recovery time courses. However, unlike the present study, multi-match congestion was shown to further exacerbate perceived fatigue ratings following the 2<sup>nd</sup> match.<sup>16</sup> The contradictions between these findings may in part be due to the contextual differences of club football, as the previous study included a club team with excessive post-match travel demands. Further, factors like club training programmes with higher post-match training loads, or limited/reduced access to recovery facilities must also be considered in domestic Australian clubs.<sup>10,17</sup> Nevertheless, within the current study exposure to acute tournament match congestion does not appear to exacerbate international footballer’s post-match fatigue and recovery responses. Although the generalisability of this finding is unknown and future research in other national teams over longer tournament periods are warranted.

Closer examination of the between-group comparison of the multi-match conditions alludes to significant variations in player perceptual ratings between Acute-Congestion and Non-Congestion microcycles despite similar load profiles. Specifically, moderate effect sizes existed for worse recovery of perceived fatigue, muscle soreness and sleep duration in Acute-Congestion. Previously, research within club-based contexts identified these parameters as the most responsive to match load during multi-match weeks, citing the repeated matches, logistical demands, and limited opportunity for recovery as justification for the observed

responses.<sup>15,16</sup> However, exposure to match load may not be the only determinant of perceived fatigue and recovery responses, as other match-related factors including match outcome and players' individual performance can influence post-match perceived fatigue, stress and sleep.<sup>29</sup> Although, the authors acknowledge these factors were not controlled for within this analysis, their contribution may only be evident up to 24h post-match.<sup>29</sup> Notwithstanding, day-to-day variances within the No-Match condition were also evident within this study. Most notably total perceived response values were worsened on MD1+1, due largely to higher (poorer) ratings in sleep duration, with some late-night fixtures and possible air travel likely compromising player's sleep patterns.<sup>30</sup>

The association between match-day load and the subsequent post-match perceived response had moderate-to-large positive correlations for all items, except psychological status. These findings align with research by Moalla and colleagues<sup>12</sup> who reported similar correlations ( $r = 0.23$ – $0.48$ ) between daily internal training load (s-RPE) and subjective ratings in professional footballers during pre- and in-season periods. Notably, perceived ratings of fatigue and muscle soreness were the most strongly associated with match loads, likely due to the high physical demands of football match-play resulting in residual post-match fatigue and tiredness.<sup>5</sup> In contrast, sleep quality and sleep duration were less sensitive to match loads, which may be a reflection of their responsiveness to contextual match factors (i.e. reduced recovery time and travel demands) rather than the match load itself. However, removing the no-match condition showed no significant correlations between any perceived response items and s-RPE match load. Thus, player perceptual responses may be responsive to the presence of match load, but is likely not sensitive to the variations/fluctuations evident within congested microcycles. Indeed, recent concerns have been raised relating to the validity and reliability of using single-item and composite scores to measure athlete's training 'response', with these measures unlikely to be influenced by only training and match load.<sup>30,31</sup>

## 293 **Conclusion**

294 Variations in perceived load, fatigue and recovery profiles between congested and non-  
295 congested microcycles were largely determined by exposure to tournament match-play. While  
296 a consistent trend of worsened subjective ratings for Acute-Congestion was observed  
297 compared to the other congestion conditions, similarities in the magnitude of player responses  
298 on MD1 and MD2 suggest that acute match congestion does not exacerbate perceived fatigue  
299 and recovery responses within the context of international football tournaments.

300

## 301 **Practical implications**

302 During international football tournaments, periods of acute-match congestion were shown to  
303 be followed by impaired states of pre-match subjective status compared to non-congested  
304 microcycles. However, the sensitivity of players' perceived total response to training and  
305 match load appears to be less evident within congested microcycles. Future research should  
306 aim to determine the sensitivity, if not validity, of the subjective metrics in relation to external  
307 load or another response measure during congested schedules.

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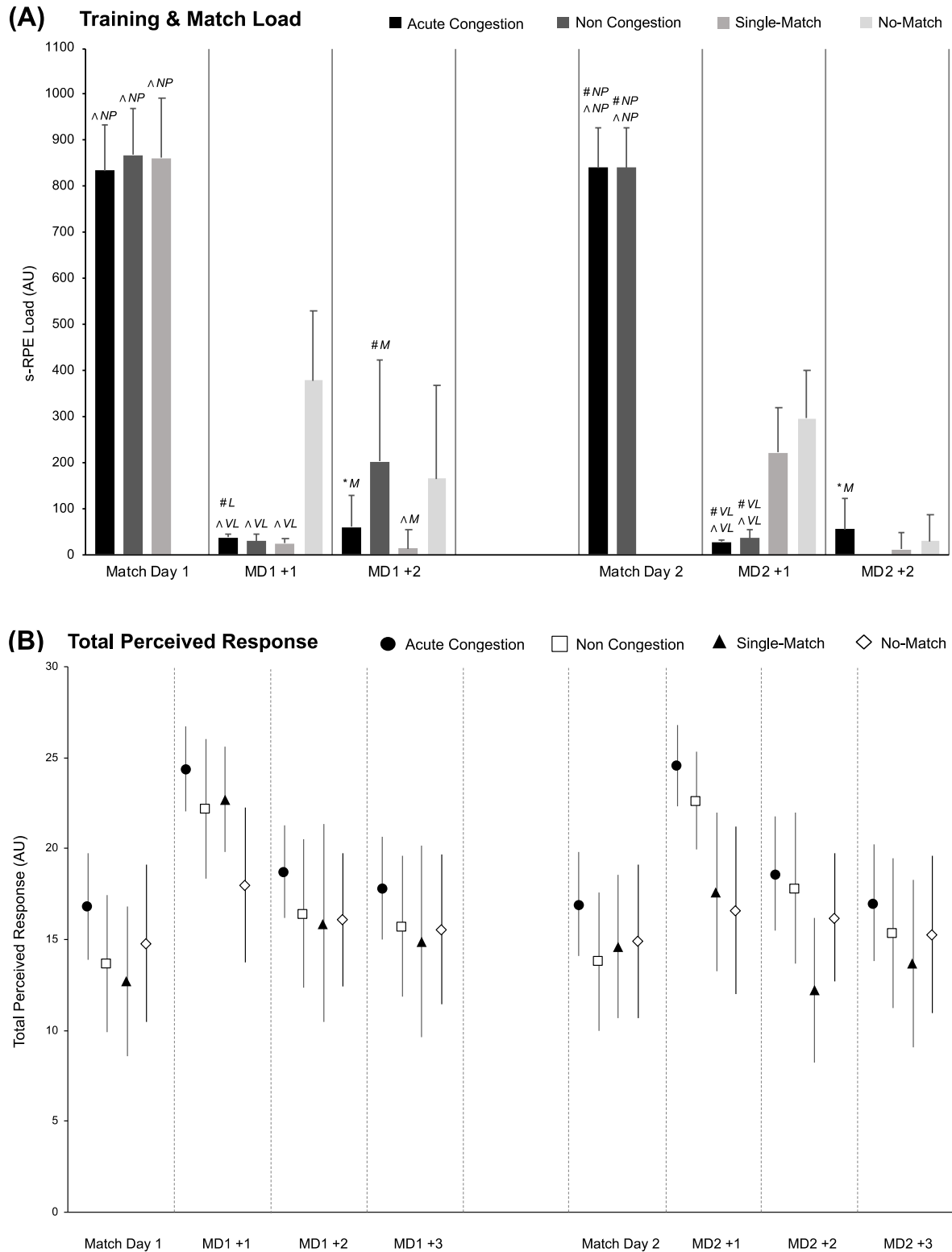
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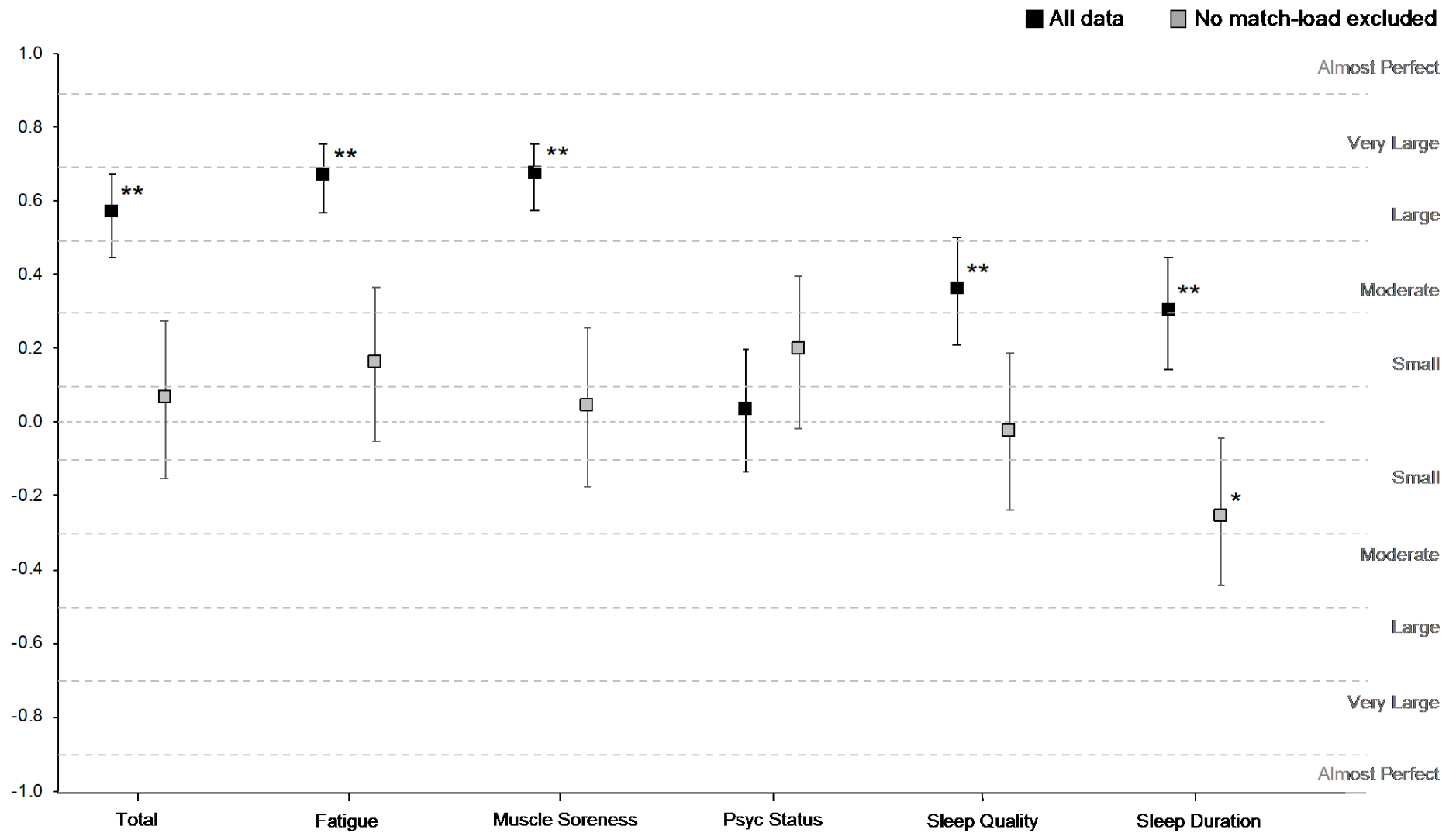
**Figure 1.** Internal (s-RPE) training/match load (Figure A) and total perceived response profiles (Figure B) for Acute Congestion, Non-Congestion, Single-Match, and No-Match microcycles. Note: Higher perceived response scores represent a worsened status. \*Denotes a significant difference from NC, #Denotes a significant difference from SM, ^Denotes a significant difference from NM;  $p \leq 0.05$ . Effect size abbreviations: *S* = small, *M* = Moderate, *L* = large, *VL* = very large, *NP* = nearly perfect.

**Table 1.** Sub-scale perceived response profiles for Acute-Congestion, Non-Congestion, Single-Match, and No-Match microcycles.

	MD1	MD1 +1	MD1 +2	MD1 +3	MD2	MD2 +1	MD2 +2	MD2 +3
<b>Fatigue</b>								
Acute-Congestion	3.4 ± 0.8	5.2 ± 0.5 <sup>^L</sup>	4.1 ± 0.8 <sup>^M</sup>	3.7 ± 0.8 <sup>^M</sup>	3.6 ± 0.6 <sup>*M #L ^M</sup>	5.2 ± 0.4 <sup>#VL ^VL</sup>	4.3 ± 0.6 <sup>#VL ^L</sup>	3.7 ± 0.8 <sup>^M</sup>
Non-Congestion	2.6 ± 1.2	4.9 ± 1.0 <sup>^L</sup>	3.7 ± 1.1	3.2 ± 1.2	2.7 ± 1.1	4.9 ± 0.9 <sup>#L ^L</sup>	4.0 ± 1.0 <sup>#L ^M</sup>	3.3 ± 1.1
Single-Match	2.3 ± 1.3	4.4 ± 1.4	3.2 ± 1.6	2.8 ± 1.4	2.2 ± 1.1	2.6 ± 1.5	1.9 ± 1.3	2.3 ± 1.4
No-Match	2.7 ± 1.3	3.1 ± 1.3	3.0 ± 1.3	2.8 ± 1.3	2.7 ± 1.2	2.8 ± 1.3	2.9 ± 1.3	2.7 ± 1.3
<b>Muscle Soreness</b>								
Acute-Congestion	3.6 ± 0.8	5.0 ± 0.7 <sup>^L</sup>	4.2 ± 0.5 <sup>^M</sup>	3.7 ± 0.8	3.6 ± 0.7 <sup>#L ^M</sup>	5.1 ± 0.4 <sup>#VL ^VL</sup>	4.4 ± 0.7 <sup>#VL ^L</sup>	3.7 ± 0.8 <sup>#L ^M</sup>
Non-Congestion	2.7 ± 1.2	4.8 ± 0.8 <sup>^L</sup>	3.9 ± 1.1	3.4 ± 1.1	2.9 ± 1.1	4.9 ± 0.8 <sup>#VL ^VL</sup>	4.2 ± 1.0 <sup>#L ^M</sup>	3.4 ± 1.2
Single-Match	2.7 ± 1.2	4.8 ± 0.8 <sup>^L</sup>	4.2 ± 1.1	3.3 ± 1.2	2.6 ± 0.8	2.7 ± 1.1	2.4 ± 1.1	2.5 ± 1.0
No-Match	3.0 ± 1.3	3.3 ± 1.2	3.2 ± 1.2	3.0 ± 1.1	2.8 ± 1.1	2.7 ± 1.3	3.2 ± 1.0	2.8 ± 1.1
<b>Psychological Status</b>								
Acute-Congestion	2.7 ± 1.0	3.1 ± 1.0	2.9 ± 1.0	2.7 ± 0.9	2.7 ± 0.9	3.2 ± 1.0	3.0 ± 0.9	2.6 ± 1.0
Non-Congestion	2.3 ± 0.8	2.5 ± 1.0	2.3 ± 0.7	2.3 ± 0.8	2.3 ± 0.7	2.6 ± 1.1	2.5 ± 0.9	2.3 ± 0.8
Single-Match	2.1 ± 1.0	2.8 ± 1.0	2.4 ± 1.1	2.2 ± 0.9	2.5 ± 1.2	2.7 ± 0.9	2.4 ± 0.9	2.3 ± 0.9
No-Match	2.5 ± 1.2	3.0 ± 1.1	1.9 ± 1.2	2.9 ± 1.2	2.7 ± 1.2	3.0 ± 1.3	3.2 ± 1.2	2.8 ± 1.1
<b>Sleep Quality</b>								
Acute-Congestion	3.2 ± 0.8	4.9 ± 0.9 <sup>^L</sup>	3.4 ± 1.0	3.4 ± 0.9	3.2 ± 0.9	4.9 ± 1.3 <sup>^L</sup>	3.1 ± 1.3	3.2 ± 0.9
Non-Congestion	2.8 ± 0.9	4.4 ± 1.2	2.9 ± 1.4	2.9 ± 1.2	2.7 ± 1.0	4.4 ± 1.0 <sup>^M</sup>	3.2 ± 1.6	2.7 ± 1.2
Single-Match	2.7 ± 0.9	4.5 ± 0.9	2.8 ± 1.6	3.0 ± 1.3	3.3 ± 1.1	4.2 ± 1.0	2.6 ± 0.9	2.8 ± 1.2
No-Match	3.0 ± 1.1	3.7 ± 1.1	3.3 ± 0.8	3.2 ± 0.9	3.2 ± 1.0	3.3 ± 1.0	3.4 ± 0.8	3.2 ± 1.0
<b>Sleep Duration</b>								
Acute-Congestion	3.9 ± 0.9 <sup>#M</sup>	6.2 ± 0.6 <sup>^M</sup>	4.1 ± 1.0	4.4 ± 0.9 <sup>^M</sup>	3.8 ± 0.9	6.1 ± 0.8 <sup>^L</sup>	3.8 ± 1.4	3.8 ± 1.0
Non-Congestion	3.2 ± 1.0	5.5 ± 1.3	3.5 ± 1.2	3.9 ± 1.4	3.2 ± 1.0	5.8 ± 0.8 <sup>^M</sup>	3.9 ± 1.6	3.6 ± 1.4
Single-Match	2.9 ± 0.9	6.2 ± 0.4 <sup>^M</sup>	3.3 ± 1.6	3.6 ± 1.4	4.0 ± 1.2	5.4 ± 1.0	2.9 ± 1.2	3.8 ± 1.2
No-Match	3.6 ± 0.8	5.0 ± 1.4	3.7 ± 0.8	3.7 ± 0.8	3.6 ± 0.8	4.7 ± 1.2	3.6 ± 1.0	3.8 ± 1.1

MD1, Match Day 1; MD2, Match Day 2

\*Denotes a significant difference from NC, #Denotes a significant difference from SM, ^Denotes a significant difference from NM;  $p \leq 0.05$ .Effect size abbreviations: <sup>S</sup> = small, <sup>M</sup> = Moderate, <sup>L</sup> = large, <sup>VL</sup> = very large



**Figure 2.** Correlation coefficients ( $r_s$ ) between s-RPE match load and post-match MD-1 perceived response items, plotted as means  $\pm$  95% CI for all data (black squares) and no-match load data excluded (grey squares).

\*correlation is significant ( $p < 0.05$ )

\*\*correlation is significant ( $p < 0.001$ )