

1 **Mathematical coupling causes spurious correlation within the conventional acute-to-**  
2 **chronic workload ratio calculations**

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## 23 INTRODUCTION

24 The monitoring of training workloads is now a much-researched topic in team sports.<sup>1</sup>  
25 Within this topic, researchers and practitioners are particularly interested in the impact of  
26 relatively short (acute) periods of higher training workloads normalised for the prior and  
27 longer-term (chronic) workloads. In recent years, a well-established approach for normalising  
28 this acute ‘spike’ to chronic load has been by calculating the “acute:chronic workload ratio”  
29 (ACWR). Both this index and chronic workload itself have been reported to be independent  
30 predictors of training-related injuries.<sup>2</sup> It has also been reported, particularly in team sports  
31 competitors, that there are associations between acute spikes in training workloads (relative  
32 to chronic workloads) and time-loss injuries.<sup>1</sup>

33 The ACWR is usually calculated as the simple ratio of recent (i.e. one-week) to  
34 longer term (i.e. four-week) training workloads.<sup>1</sup> While it is important for the numerator and  
35 denominator of any ratio to be correlated only through biological mechanisms,<sup>3</sup> one aspect of  
36 the ACWR calculation is that the acute workload also constitutes a substantial part of the  
37 chronic workload.<sup>4</sup> This “*mathematical coupling*” between two variables,<sup>5</sup> also referred to as  
38 “*relating a part to the whole*”,<sup>6</sup> is unusual and raises the possibility that research inferences  
39 and athlete monitoring might be compromised by resulting spurious correlations.<sup>3</sup> A spurious  
40 correlation is one which exists between two variables irrespective of any true  
41 biological/physiological association between those variables.<sup>3,5</sup>

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## 43 MATHEMATICAL COUPLING IN THE ACWR CALCULATION

44 Irrespective of different data smoothing approaches over a 28-day period,<sup>7</sup> the  
45 conventional calculation of the ACWR is ultimately:

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$$47 \quad ACWR = \frac{A}{0.25 \cdot (W1 + W2 + W3 + A)}$$

48 where A is the 7-day acute workload and hypothetical W1, W2 and W3 are the preceding 7-  
49 day workloads, respectively.<sup>1,4</sup> Given the conceptual definition of acute and chronic workload  
50 variables<sup>4</sup> we hypothesised that “mathematical coupling” might exist, leading to a spurious  
51 correlation between acute and chronic workload estimates.<sup>3</sup>

52 To test our hypothesis with adequate statistical precision, we generated data to  
53 simulate four 7-day periods of high-speed distance data reported in a recent study involving  
54 elite Australian footballers<sup>2</sup> for a hypothetical squad of 1000 players (Supplementary file).  
55 Each of the four sets of data was randomly generated and was completely independent from  
56 the other datasets. The most recent 7-day period was designated as the acute period (A),  
57 while the 28-day period defining chronic workload was calculated as a conventional rolling  
58 average. The mean±SD high-speed distance for W1, W2 and W3 and A were 2021±889 m,  
59 1977±880 m, 1968±860 m and 2035 ±901 m, respectively. None of the preceding 7-day  
60 datasets were found to be substantially correlated with A ( $r < 0.06$ ). However, as  
61 demonstrated in Figure 1, there was a moderate-to-large, positive correlation between the  
62 calculated chronic and acute workload data;  $r = 0.52$  (95%CI: 0.47 to 0.56). If A was not  
63 included in the calculation of C, then the correlation between A and C was, as expected, close  
64 to zero;  $r = 0.01$  (95%CI: -0.05 to 0.07).

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66 *Figure 1 about here*

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68 The moderate-to-large but spurious (false) correlation between the acute and chronic  
69 workload variables substantiated the presence of mathematical coupling, since the acute  
70 workload represents a term in the calculation of the denominator in the ACWR.<sup>3</sup> Any  
71 functions that are designed to quantify the association between acute and chronic workload  
72 variables must be mathematically distinct from each other and not naturally associated if any

73 true physiological explanations or likelihood of injury are attempted to be researched.<sup>3</sup>  
74 Accordingly, the mathematical coupling issues we observed could also affect the chronic  
75 workload variance and, crucially, its physiological range of measurements.<sup>3</sup> In our simulated  
76 data, the SD for chronic high-speed distance (with the acute data period included) was  $\pm 439$   
77 m (data range: 654 to 3469 m). Nevertheless, following removal of the acute period data from  
78 the calculation of the chronic period distance, the SD was a higher  $\pm 499$  m (data range: 541  
79 to 3553 m). Furthermore, the formulation of rolling averages might also influence the  
80 observed SD.<sup>8</sup> Therefore, and as expected, inclusion of the acute data in the calculation  
81 artifactually reduced the between-athlete variability in chronic workload.

82         The mathematical coupling issue can also alter the ACWR itself. For example, with  
83 an acute distance of 2375 m, the chronic distance can be calculated conventionally to be 1639  
84 m. But this value without mathematical coupling should really be 1393 m. The respective  
85 ACWRs are 1.45 with the acute period included in the chronic calculation vs 1.71 when the  
86 acute data are not included in the chronic calculation. Therefore, the traditional mathematical  
87 definition of the chronic workload term in the ACWR protocol also appears to limit a valid  
88 and unbiased interpretation of the observed ACWR estimates.

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## 90 **CONCLUSIONS**

91         Collectively, our findings have demonstrated that the numerator and denominator in  
92 the ACWR are mathematically coupled and, therefore, spuriously correlated. The simplest  
93 solution is not to include acute workload periods in the calculation of chronic workload if the  
94 workload-injury aetiological relationship, grounded on the magnitude of the ACWR, is to be  
95 interpreted accurately.

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98 **Contributors**

99 LL and GA developed the article concept. All authors contributed to write, provide feedback,  
100 and revise the manuscript.

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102 **Competing interests**

103 None declared

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105 **Provenance and peer review**

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## FIGURE LEGENDS

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129 **Figure 1. The spurious correlation between the simulated acute phase data and the**  
130 **chronic phase data.** Although the four weeks of data were uncorrelated with each other, this  
131 correlation is explained by the fact that the acute phase data is part of the calculation of the  
132 chronic phase data leading to mathematical coupling. This spurious correlation will be  
133 present irrespective of any true physiological association between acute and chronic  
134 workloads, leading to biased inferences.

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