

1 Type of the Paper (Conference Paper)

2 **The Compound Score in elite road cycling**

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11 **1. Introduction**

12 Elite road cycling is characterized by racing  
13 over varied terrain, ranging from flat races to  
14 extremely mountainous terrain<sup>1,2</sup>.  
15 Researchers have frequently attempted to  
16 quantify the performance characteristics of  
17 cyclists to predict race success based on  
18 external and internal load metrics including  
19 power output, heart rate and speed<sup>3-6</sup>. Recent  
20 research reported a strong relationship  
21 between the power profile and race  
22 performance<sup>7,8</sup>. However, to date, there is still  
23 an ongoing debate whether absolute power  
24 output; a mass exponent, or relative power;  
25 power output normalized to body mass, is  
26 more advantageous. For this reason, the  
27 current study used both absolute and relative  
28 power output to calculate a compound score  
29 to investigate its predictive ability for race  
30 performance.

31 **2. Materials and Methods**

32 Power output data were recorded from  
33 power meter system (SRAM Red, Quarq,  
34 Spearfish, South Dakota, USA) fitted to the  
35 participants bicycle (Revelator Alto Elite,  
36 KTM Fahrrad GmbH, Mattighofen, Austria)  
37 during training and racing in a competitive  
38 racing session. Body mass (Kern DS 150k1,  
39 Kern & Sohn, Germany) was recorded in  
40 conjunction with racing events. Data from  
41 training and racing data were analyzed  
42 (WKO5, Trainingpeaks LLC, US) together  
43 with a novel adaptation of these data - the

44 compound score. In a second step absolute,  
45 relative power output and the compound  
46 score were compared to performances in  
47 races to assess whether individual variables  
48 were correlated with performance, and to  
49 derive positive and negative predictive  
50 values.

51 The Compound Score can be calculated as  
52 follows:

53  
54 
$$\text{Compound Score [W}^2\cdot\text{kg}^{-1}] = \text{absolute power}$$
  
55 
$$\text{output [W]} \times \text{relative power output [W}\cdot\text{kg}^{-1}]$$
  
56 
$$\text{Equation (1)}$$

58 Race performances during the season for  
59 each participant were screened to select the  
60 best 3 single day race results. Results were  
61 log transformed and weighted accordingly as  
62 follows:

63 **Table 1.** represents weighting factors according to  
64 single day race categories

Single Day	
Cat	Weighting
1.1	2
1.2	1.5
1.2 U23	1.1
NC	1
1.2 NC	0.8

65 Cat – category, NC – nation cup

66 Subjects – Thirty male U23 professional  
67 cyclists participated in the study (age,  
68 20.1±1.1, body mass 69 ± 6.9 kg, height 182.6  
69 ± 6.2cm) All participants provided informed  
70 written consent and were active members of



71 a UCI Continental team during the cycling  
 72 season(s) analyzed.

73

74 Statistical Analysis — All values are  
 75 expressed as mean ± standard deviation and  
 76 or mean difference (MD). A Pearson product  
 77 correlation was used to investigate the  
 78 relationship between 5-min MMP, ( $W, W.kg^{-1}$ ),  
 79 compound score of 5-min and the best  
 80 single day result score. The correlation  
 81 coefficient was interpreted according to  
 82 Hopkins<sup>9</sup> for a small (<.3), medium (.3-.5) or  
 83 large (>.5) effect. The performance threshold  
 84 was calculated as the ratio from the true to  
 85 false observations, which were below or  
 86 above the corresponding cut offs relating to  
 87 5-min MMP, ( $W, W.kg^{-1}$ ) and compound  
 88 score. All statistical analyses were completed  
 89 using GraphPad Prism (version 8.0.0 for Mac  
 90 OS, GraphPad Software, San Diego, USA).  
 91 The alpha level of statistical significance was  
 92 set as  $p > .05$  two tailed.

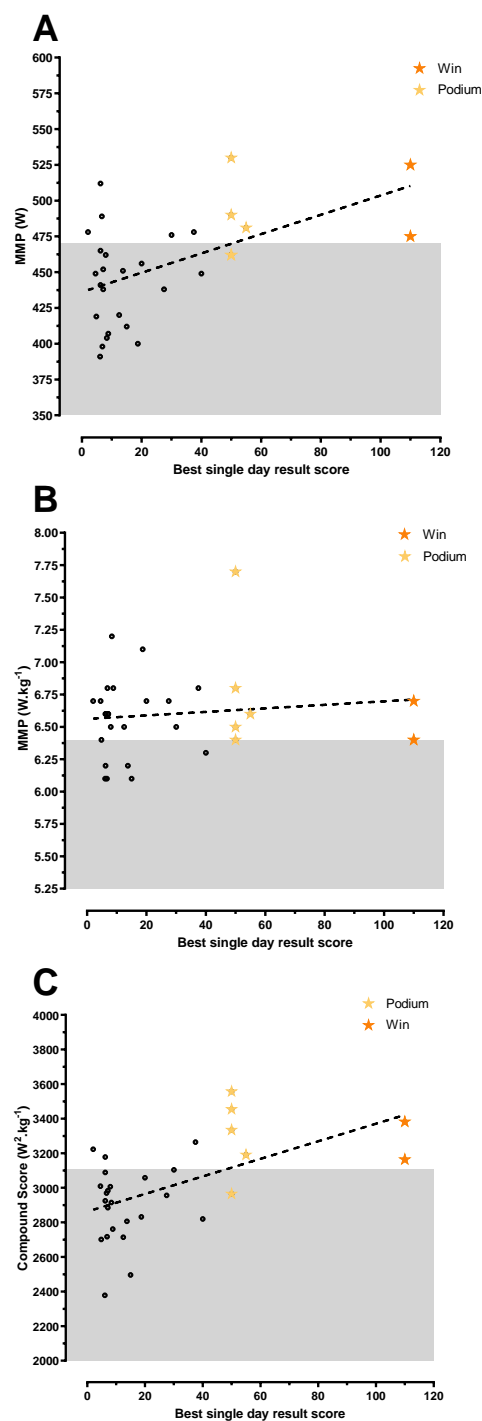
### 93 3. Results

94 **Table 1.** demonstrates the participants'  
 95 descriptive performance characteristics.

5-min MMP (W)	5-min MMP ( $W.kg^{-1}$ )	Compound Score ( $W^2.kg^{-1}$ )
$445 \pm 36$	$6.6 \pm 0.3$	$2995 \pm 264$

96 5-min MMP – 5 minute mean maximum power

97 Absolute MMP ( $r=.52, p=.003$ ) and the  
 98 compound score ( $r=.54, p=.002$ ) significantly  
 99 correlated with the best single day result  
 100 score, while relative MMP did not ( $r=.11,$   
 101  $p=.550$ ). Positive/negative performance  
 102 thresholds were  $>470W, 50.0/90.0\%$ ; for  
 103 absolute MMP,  $>6.4 W.kg^{-1}, 20.8/50.0\%$  for  
 104 relative MMP and  $>3110 W^2.kg^{-1}, 66.7/95.2\%$   
 105 for the compound score respectively – see  
 106 figure 1.



**Figure 1.** Illustrates the relationship between absolute (A); relative (B) mean maximum power (MMP); compound score (C) and the best single day result score. The grey shaded area represents the performance threshold for the variable used to predict a race podium or win.

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### 116 4. Discussion

117 In keeping with our hypothesis that both  
 118 a high absolute power output as well as a  
 119 high relative power output are important in  
 120 determining performance, we have  
 121 demonstrated that the product of these two  
 122 variables has a greater correlation with and is  
 123 able to predict a successful race outcome to a  
 124 greater extent than either variable alone. The  
 125 two greatest forces a cyclist is required to  
 126 overcome are gravitational force and drag.  
 127 The former requires a high relative power  
 128 output while the latter requires absolute  
 129 power. As relative power output scales  
 130 inversely to mass and absolute power output  
 131 scales proportionally with mass, these two  
 132 variables represent a diverging set of  
 133 performance characteristics relative to the  
 134 mass of the rider. As such, there may be a  
 135 mass at which cyclists exhibits an optimal  
 136 balance between these two characteristics to  
 137 achieve the highest performance  
 138 characteristics. The compound score seeks to  
 139 provide a variable with which the balance of  
 140 these diverging performance variables can be  
 141 measured. We have demonstrated that for  
 142 U/23 professional cycling, a compound score  
 143 of 3110 W<sup>2</sup>.kg<sup>-1</sup> has a 66.7% positive  
 144 predictive value for the achievement of a  
 145 podium or race win result. Conversely, a  
 146 compound score less than 3110 W<sup>2</sup>.kg<sup>-1</sup> is  
 147 associated with a 95.2% negative predictive  
 148 score. i.e., a compound score below this value  
 149 is associated with only 4.8% likelihood of a  
 150 race podium result. To our knowledge, the  
 151 compound score is able to measure  
 152 performance characteristics for U23 one day  
 153 racing success. Further research is required to  
 154 assess whether the compound score is able to  
 155 predict stage race success or whether other  
 156 factors such as the power profile<sup>8</sup> or fatigue  
 157 resistance<sup>10-12</sup> provide greater insight.

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163 **Conflicts of Interest:** The authors declare no  
 164 conflict of interest.

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