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The cycling power profile characteristics of national level junior triathletes

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25 **ABSTRACT**

1 With the draft-legal rule recently introduced to junior triathlon competition, it has become
2 difficult to assess cycling performance through race results. Therefore, this study assessed the
3 cycling power profile characteristics of national level junior triathletes to assist with physical
4 assessment and program design. Thirteen male (17.0 ± 1.0 yr) and eleven female (17.2 ± 1.3
5 yr) national level junior triathletes completed a cycling power profile that consisted of maximal
6 intervals that lasted 6, 15, 30, 60, 240 and 600 seconds in duration. Each power profile was
7 completed on a LeMond ergometer using the subject's own bicycle, with power output and
8 cadence recorded for all intervals. Mean power output values for males (783 ± 134 , 768 ± 118 ,
9 609 ± 101 , 470 ± 65 , 323 ± 38 , 287 ± 34 W) were significantly ($P < 0.05$) higher than females
10 (554 ± 92 , 510 ± 89 , 437 ± 75 , 349 ± 56 , 248 ± 39 , 214 ± 37 W) across all intervals, respectively.
11 Peak power output values for males across the 6 and 15 second intervals (1011 ± 178 and 962
12 ± 170 W) were also significantly higher than for females (674 ± 116 and 624 ± 114 W),
13 respectively ($P < 0.05$). Developing junior triathletes should aim to increase their capacity
14 across the power profile above the mean values listed. Athletes should further aim to have
15 power outputs equal to that of the best performers and beyond to ensure that they can meet the
16 demands of any competition situation.

17

18 **Keywords:** triathlon, youth, draft-legal, coaching, testing

19

1 INTRODUCTION

2 Triathlon is a multidisciplinary sport encompassing the sequential completion of swimming,
3 cycling and running stages. In elite senior and junior competition, racing is classed as ‘draft-
4 legal’, permitting athletes to closely follow one another (i.e. drafting) during the cycling stage
5 to reduce drag forces (2, 11). While drafting may also be beneficial during the swimming and
6 running stages, it has particular importance during the cycling stage due to the increased wind
7 resistance creating greater drag at high speeds (12). Specifically, drafting behind small (i.e. 1-
8 4 riders) and large (i.e. 8 or more riders) groups of cyclists has been shown to reduce the oxygen
9 consumption requirement to sustain a given speed by as much as $27 \pm 7\%$ and $39 \pm 6\%$,
10 respectively (5). Hence, drafting allows individual competitors to alternate between higher
11 intensity efforts whilst leading the group or making a breakaway manoeuvre, with interspersed
12 lower intensity efforts when drafting to conserve energy. A study of male international
13 Olympic distance triathlon competition revealed that 34 ± 14 high intensity efforts (>600 W)
14 were performed during the cycling stage and 18% of total cycling time exceeded maximal
15 aerobic power (3), highlighting the intermittent demands of the race. Hence, the tactical nature
16 of drafting transforms the demands of the cycling stage into a high-intensity, intermittent
17 activity.

18

19 Due to the tactical nature of the draft-legal format, the cycling performance of opponent
20 triathletes during such competitions (i.e. their maximal performance over various durations) is
21 difficult to assess. Performance in the swimming and running stages can be inferred from race
22 times due to these stages more closely reflecting an individual time trial. However, in the
23 cycling stage, athletes take advantage of the draft effect and ride together in groups, which
24 means that they often finish with the same time (2). Also, many athletes will attempt to
25 minimise power output during the cycling stage in order to conserve energy prior to the running

1 stage (2, 8). Therefore, the optimal way to assess the maximal cycling capability of an athlete
2 over various durations is through controlled laboratory testing.

3
4 Current laboratory-based research on cycling in triathlon has focused on assessing maximal
5 aerobic capacity using incremental test protocols, with values as high as $74.3 \pm 4.3 \text{ mL}\cdot\text{kg}^{-1}\cdot\text{min}^{-1}$
6 reported for elite senior competitors (6). Further, maximal aerobic power values of 385-
7 389 W have been reported for senior elite triathletes (4, 6). The application of such data for
8 draft-legal races is questionable considering that the high intensity, intermittent profile of the
9 draft-legal format requires the assessment of a triathlete's complete aerobic and anaerobic
10 capacities across various durations (3, 8). The cycling power profile is a reliable performance
11 test incorporating maximal self-paced intervals of 6-600 seconds in duration (8) and it has
12 recently been demonstrated to predict road cycling performance (8). It has also been
13 recommended by the Australian Institute of Sport as a useful cycling test protocol for triathletes
14 (13) and as a result it has been adopted by Australian state-level junior representative triathlon
15 squads. As such, this test has become important for physical assessment and program design
16 for these junior athletes, however, no normative data currently exists for this population, which
17 would likely assist coaches and athletes with their interpretation of test results. Therefore the
18 purpose of this study was to describe the laboratory power profile results of junior male and
19 female triathletes competing at the national level.

20

21 **METHODS**

22 **Experimental Approach to the problem**

23 This descriptive study measured the power profile performance of national level junior
24 triathletes in a standardised laboratory test consisting of six maximal self-paced intervals (6,
25 15, 30, 60, 240 and 600 s in duration) with periods of active recovery (174, 225, 330, 480 and

1 600 s in duration) as described previously (8). All cycling was completed on each subjects's
2 own personal road bicycle that was attached to a LeMond Revolution cycle ergometer
3 (LeMond Fitness Inc., Woodinville, Washington, USA). The LeMond Revolution takes the
4 place of the rear wheel, using the bicycle's normal drivetrain to adjust resistance, which allows
5 the use of equipment and bicycle geometry that is specific to each individual. Power output
6 obtained from the LeMond Power Pilot (LeMond Fitness Inc., Woodinville, Washington,
7 USA) has previously been validated against the SRM power meter with the level of agreement
8 considered acceptable (7). Data was collected during training camps leading into competition
9 when the athletes were close to their peak condition.

10

11 **Subjects**

12 Thirteen male (age: 17.0 ± 1.0 yr, stature: 176.6 ± 5.7 cm, body mass: 65.8 ± 7.1 kg, sum of 7
13 skinfolds: 49.4 ± 10.2 mm, body fat: $8.7 \pm 1.7\%$) and eleven female (age: 17.2 ± 1.3 yr, stature:
14 166.8 ± 7.9 cm, body mass: 57.5 ± 7.7 kg, sum of 7 skinfolds: 76.5 ± 15.5 mm, body fat: 16.8
15 $\pm 3.9\%$) national level junior triathletes volunteered for the study. Inclusion criteria stipulated
16 that subjects must be aged 16-19 years and currently competing in the Australian National
17 Junior Triathlon Series over the sprint distance (i.e. 750 m swim, 20 km cycle, 5 km run). All
18 subjects were familiar with riding on a cycle ergometer. All subjects and their guardians
19 provided written informed consent prior to testing. An institutional ethics committee granted
20 approval for the project (XXX H-2011-0350).

21

22 **Procedures**

23 An anthropometric profile was obtained from each participant consisting of stature (217
24 Stadiometer, Seca, Birmingham, United Kingdom), body mass (DS-530 electronic scales,
25 Wedderburn, Sydney, Australia) and skinfold thickness at seven sites (Harpden Calipers,

1 Baty International, West Sussex, United Kingdom). The seven sites included bicep, tricep,
2 subscapular, supraspinalae, abdominal, quadriceps and medial calf and these sites were
3 summed to form the sum of 7 skinfolds (X_1). Body density was calculated with specific
4 regression equations for male (14) and female (15) Australian athletes as per below (where X_2
5 = the sum of 6 skinfolds as above minus the bicep). Percent body fat was also estimated via the
6 equation below (9).

$$7 \text{ Male Body Density (14)} = 1.0988 - 0.0004(X_1)$$

$$8 \text{ Female Body Density (15)} = 1.20953 - 0.08294(\text{Log}_{10}X_2)$$

$$9 \text{ \% Body Fat (9)} = [4.95/\text{Body Density} - 4.5] \times 100$$

10
11
12 For 24 hours prior to the power profile, caffeine and high intensity exercise were not permitted
13 and the athletes were instructed to consume their usual pre-race diet. The participants
14 performed a standardised 10 min warm-up that consisted of riding between 100-200 W, as well
15 as three six second intervals at 70, 80 and 90% of their perceived maximal intensity,
16 respectively. The power profile test commenced two minutes later and all intervals began from
17 a rolling start between 70-80 $\text{r} \cdot \text{min}^{-1}$. Verbal encouragement was provided during the intervals
18 and participants were instructed to self select and adjust their gear ratio at any time to produce
19 their best performance over each interval. The athletes were also instructed that the shorter
20 intervals (6-15 s) were a maximal sprint while the longer intervals (30-600 seconds) required
21 a self-selected pacing strategy to produce the maximal mean power. During active recovery,
22 cyclists were instructed to pedal at a power output of <100 W. A 50 centimetre fan was placed
23 1 metre in front of the participant and provided a wind speed of 8 $\text{m} \cdot \text{s}^{-1}$ to simulate the
24 convective cooling of outdoor conditions and tepid water (20-23°C) was ingested *ad libitum* as
25 recommended (10).

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Measures

Power output and cadence were recorded at a frequency of 1 Hz using a LeMond Power Pilot. The first second of data obtained in the 6 second intervals was not included in the data analysis as per previous research (8). Heart rate was recorded with a Garmin Forerunner 910XT heart rate monitor wrist watch and chest strap (Garmin Ltd., Canton of Schaffhausen, Switzerland). All data was downloaded post-test and arranged in Microsoft Excel (Microsoft Corporation™, Redmond, WA, USA) before further analysis. Power output data were also divided by the participant's body mass to calculate relative values.

Statistical Analyses

The data were examined for assumptions of normality using the Kolmogorov-Smirnov test and visually inspected through histograms and box plots. A two-way repeated measures ANOVA was used to determine the main effects of sex on power output, cadence and heart rate for each interval where it was measured. *Post hoc* comparisons with Bonferonni adjustment were used to identify any significant differences. All statistical analysis were conducted using SPSS software V22.0 (IBM Corporation, Somers, NY, USA). Power curves were plotted for each athlete and group means using Microsoft Excel's built-in power function ($R^2 > 0.94$ for all power curves) and a 'best performer' for both sexes was identified as the athlete who achieved the highest power output across all interval durations in the power profile itself and does not necessarily reflect the best performing triathlete in competition.

RESULTS

The descriptive statistics for mean power output measures of the group and the best performer across the power profile are presented in Table 1. All mean power outputs reported were

1 significantly higher in males than females for both absolute and relative measures ($P<0.05$).

2 Power curves of the group means and best performing male and female athlete across the power
3 profile tests are presented in Figure 1.

4

5 ***Insert Table 1 Here***

6 ***Insert Figure 1 Here***

7

8 The descriptive statistics for peak power output measures of the group and the best performer
9 across the 6 second and 15 second intervals are presented in Table 2. These peak power outputs
10 were both significantly higher in males when compared to females for both absolute and
11 relative measures ($P<0.05$).

12

13 ***Insert Table 2 Here***

14

15 Mean and peak cadence measures of the group and best performer are presented in Table 3.
16 Mean cadence measures were significantly higher in males when compared to females across
17 the 15 and 30 second intervals ($P<0.05$). Peak cadence measures were significantly higher in
18 males when compared to females across the 6 and 15 second intervals ($P<0.05$). There were
19 no significant differences in cadences across any other interval ($P>0.05$).

20

21 ***Insert Table 3 Here***

22

23 Mean heart rates across the 240 and 600 s intervals were 172 ± 8 beats·min⁻¹ and 179 ± 6
24 beats·min⁻¹ as well as 174 ± 7 beats·min⁻¹ and 178 ± 5 beats·min⁻¹ for males and females,

1 respectively. No significant differences were observed between sexes for the heart rate
2 measures ($P>0.05$).

3

4 **DISCUSSION**

5 This investigation has provided a novel insight into the cycling capacities of national level
6 junior triathletes. This information is useful for a number of purposes including the preparation
7 of athletes, monitoring changes in performance and talent identification. Such data provides a
8 set of normative values for regular cycle-based testing, which can also help to identify specific
9 strengths and weaknesses to benefit training prescription. Overall, the males outperformed the
10 females, even when corrected for differences in body mass, although the gap between relative
11 data for males and females was somewhat reduced. Further, males and females employed
12 significantly different cadences for the intervals shorter than 60 seconds duration, however
13 both cadences and physiological intensities were similar for the longer duration intervals.

14

15 The power output requirements of the cycling stage within draft-legal junior triathlon are
16 highly variable, with the employed race tactics depending on a wide range of variables (2). In
17 addition, each course is highly variable, consisting of an entirely different circuit profile.
18 Therefore it is not adequate to prepare for such a race in this competition by simulating a
19 previous race in training (i.e. with the aid of performance times or race power outputs through
20 power meter analysis). Instead, developing junior triathletes should aim to be physically
21 superior by improving their capability to produce power across both aerobic and anaerobic
22 intervals (8), which is of high importance to draft-legal triathlon racing (3). The current study
23 described the mean cycling power outputs of junior triathletes in the power profile, but also
24 highlighted the power outputs of the best performer for both sexes. Therefore, the current data
25 should be used as a set of normative values for regular cycle-based testing in these developing

1 athletes. Developing junior triathletes and their coaches should aim initially to have power
2 outputs similar to the group mean. Secondly, athletes should aim to produce power outputs
3 equal to that of the best performer and beyond, which would ensure that they can meet the
4 demands of any competition situation and a greater opportunity for successful performance.

5

6 The use of the power profile test combined with the the data in the current study may help an
7 athlete to identify specific weaknesses in their cycling ability. Such an example may be where
8 an athlete performs well relative to their peers in the longer intervals but does not possess the
9 anaerobic power to perform well in the short duration intervals. This result would highlight the
10 need for more maximal sprint training and perhaps resistance training exercises which also
11 serves to improve cycling sprint performance (16). Another advantage of regular power profile
12 testing is that the results can be useful for a coach to construct an informed training program
13 for an athlete in relation to their current level of fitness.

14

15 Along with a set of normative values for athletes and coaches to utilise, this study provides
16 normative cycling power functions (see equations in Figure 1) for high performing junior
17 triathletes. These power functions have a useful application for training and performance
18 testing and have not previously been reported for such a cohort. Importantly, the power
19 function allows for estimation of power outputs across any duration not explicitly assessed
20 within the test protocol or for individuals that have not undertaken a power profile. By simply
21 inserting the 'x' value of the duration of interest, the power functions provided can be used to
22 estimate normative maximal mean power output across any duration between 5–600 seconds.
23 Data may also be extrapolated beyond these limits if desired, for example, comparisons of
24 functional threshold power across 20 or 60 minutes (1) would require insertion of an 'x' value
25 of 1200 or 3600, respectively. However, it should be noted that estimates may become

1 increasingly inaccurate for durations that lie further from the explicitly measured 5-600 second
2 efforts of the power profile. Nevertheless, such estimates have strong implications for coaches
3 who may be limited for time within training camps and cannot conduct a power profile assessment
4 for 50 minutes with each individual athlete. Instead, the coach may choose several efforts of
5 any duration and compare these to the normative power functions ($W \cdot kg^{-1}$) established in the
6 current study. Coaches and athletes also have the option to compare recordings from their
7 mobile power meters during field-based training and/or during races, with the normative power
8 functions established in this study.

9
10 The power outputs were significantly higher in males compared to females and these
11 differences still existed after adjustments for body mass. Interestingly, mean and peak cadences
12 were significantly lower in females compared to males for most intervals lasting less than 60
13 seconds. Considering gears were able to be freely selected by the athletes, this suggests that
14 the females preferred to perform shorter intervals at a lower cadence compared to the males. It
15 is difficult to speculate if the males would have performed better in a gear with more resistance,
16 or if the females would have performed better in a gear with less resistance. In contrast, males
17 and females chose a similar cadence in all of the intervals lasting 60 seconds or longer. Also,
18 mean heart rates were similar between the sexes across the longer duration efforts, suggesting
19 both sexes self-selected similar relative cycling intensities.

20
21 An important limitation of this study was that the study population consisted of only one fifth
22 of the triathletes competing in the Australian National Junior Triathlon Series. Indeed, a larger
23 sample size would make for a stronger set of normative data. Nevertheless, the current study
24 contained a broad spectrum of athletes, including the complete squad of two state triathlon
25 bodies. The study also includes both males and females who have gone on to compete in the

1 under 23 world triathlon championships and the senior elite category of the International
2 Triathlon Union World Triathlon Series. Hence, coaches can have confidence that the data
3 presented on the best performing athletes were of a high standard, however, there may be better
4 performing athletes who could not be included in this study. Another limitation of the study
5 was that the power profile protocol measured the power outputs from a rested state, rather than
6 a fatigued state, which would be more specific to a triathlon scenario. The ability to perform
7 anaerobic efforts under fatigue would be another useful indication of a draft-legal triathlete's
8 cycling ability.

9

10 **PRACTICAL APPLICATIONS**

11 The data described herein can be used as a set of normative values and normative power
12 functions for developing elite junior triathletes with the goal to perform well in draft-legal
13 competitions. With both the mean and best performing male and female power outputs and
14 resultant power functions clearly defined across the power profile, athletes can use these values
15 and/or equations as a training goal, or to help them identify their strengths and weaknesses
16 relative to their peers, which will be useful to inform training prescription. Overall, it allows
17 informed, evidence based decisions to be made by technical and conditioning coaches in regard
18 to the interpretation of cycling assessment and the cycling program design of national level
19 junior triathletes.

20

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22 would like to acknowledge the subjects for their contribution to the study.

23

24 **Conflict of Interest:** There is no conflict of interest pertaining to the published data.

25

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16

Table 1. Mean power output measures of the group mean and the best performer expressed in both absolute and relative terms.

| | Interval (s) | Group (W) | Best (W) | Group (W·kg⁻¹) | Best (W·kg⁻¹) |
|----------------|---------------------|------------------|-----------------|----------------------------------|---------------------------------|
| <i>Males</i> | 6 | 783 ± 134 | 1000 | 11.9 ± 1.9 | 15.7 |
| | 15 | 768 ± 118 | 920 | 11.7 ± 1.4 | 14.5 |
| | 30 | 609 ± 101 | 761 | 9.2 ± 1.1 | 12.0 |
| | 60 | 470 ± 65 | 519 | 7.2 ± 0.8 | 8.2 |
| | 240 | 323 ± 38 | 333 | 4.9 ± 0.4 | 5.2 |
| | 600 | 287 ± 34 | 321 | 4.4 ± 0.4 | 5.0 |
| <i>Females</i> | 6 | 554 ± 92* | 697 | 9.7 ± 1.2* | 10.8 |
| | 15 | 510 ± 89* | 654 | 8.9 ± 1.1* | 10.1 |
| | 30 | 437 ± 75* | 550 | 7.6 ± 0.9* | 8.5 |
| | 60 | 349 ± 56* | 455 | 6.1 ± 0.8* | 7.0 |
| | 240 | 248 ± 39* | 302 | 4.4 ± 0.7* | 4.7 |
| | 600 | 214 ± 37* | 271 | 3.8 ± 0.6* | 4.2 |

Data is presented as mean ± standard deviation. s = seconds, W = watts, W·kg⁻¹ = watts per kilogram of body mass. *Significantly ($P < 0.05$) lower than males for respective interval duration.

Table 2. Peak power output measures of the group mean and best performer expressed in both absolute and relative terms.

| | Interval (s) | Group (W) | Best (W) | Group (W·kg⁻¹) | Best (W·kg⁻¹) |
|----------------|---------------------|------------------|-----------------|----------------------------------|---------------------------------|
| <i>Males</i> | 6 | 1011 ± 178 | 1346 | 15.3 ± 1.9 | 19.3 |
| | 15 | 962 ± 170 | 1234 | 14.6 ± 2.1 | 17.7 |
| <i>Females</i> | 6 | 674 ± 116* | 864 | 11.8 ± 1.6* | 13.4 |
| | 15 | 624 ± 114* | 796 | 10.9 ± 1.4* | 12.3 |

Data is presented as mean ± standard deviation. s = seconds, W = watts, W·kg⁻¹ = watts per kilogram of body mass. *Significantly ($P < 0.05$) lower than males for respective interval duration.

Table 3. Mean and peak cadence measures of the group mean and best performer.

| | Interval (s) | Mean: Group (r·min⁻¹) | Mean: Best (r·min⁻¹) | Peak: Group (r·min⁻¹) | Peak: Best (r·min⁻¹) |
|----------------|---------------------|---|--|---|--|
| <i>Males</i> | 6 | 100 ± 9 | 92 | 118 ± 11 | 124 |
| | 15 | 112 ± 12 | 110 | 122 ± 17 | 161 |
| | 30 | 113 ± 10 | 113 | | |
| | 60 | 109 ± 11 | 113 | | |
| | 240 | 103 ± 11 | 113 | | |
| | 600 | 98 ± 10 | 101 | | |
| <i>Females</i> | 6 | 93 ± 11 | 98 | 108 ± 10* | 116 |
| | 15 | 103 ± 8* | 110 | 110 ± 10* | 119 |
| | 30 | 102 ± 9* | 108 | | |
| | 60 | 103 ± 5 | 103 | | |
| | 240 | 99 ± 7 | 100 | | |
| | 600 | 99 ± 6 | 97 | | |

Data is presented as mean ± standard deviation. r·min⁻¹ = revolutions per minute, s = seconds. *Significantly ($P < 0.05$) lower than males for respective interval duration. Peak cadence was not considered to be of relevance for intervals of >15 seconds.

Figure Captions

Figure 1: Power curves and power functions of the group means and best performing male and female athlete across the power profile tests.