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# Pacing behavior development in adolescent swimmers: a large-scale longitudinal data analysis.

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#### Abstract

**Purpose**: Use a large-scale longitudinal design to investigate the development of the distribution of effort (e.g., pacing) in adolescent swimmers, specifically disentangling the effects of age and experience and differentiating between performance levels in adulthood.

**Methods:** Season best times and 50m split times of 100m and 200m freestyle swimmers from five continents were gathered between 2000 and 2021. Included swimmers competed in a minimum of three seasons between 12-24 years old ( $5.3\pm1.9$  seasons) and were categorized by performance level in adulthood (elite, sub-elite, high-competitive) (100m: n=3498, 47% female; 200m: n=2230, 56% female). Multilevel models in which repeated measures (level 1) were nested within individual swimmers (level 2) were estimated to test the effects of age, race experience, and adult performance level on the percentage of total race time spent in each 50m section (p<0.05).

**Results:** In the 100m, male swimmers develop a relatively faster first 50m when becoming older. This behavior also distinguishes elite from high-competitive swimmers. No such effects were found for female swimmers. Conversely, more experienced male and female swimmers exhibit a slower initial 50m. With age and race experience, swimmers develop a more even velocity distribution in the 200m. Adolescent swimmers reaching the elite level adopt a more even behavior compared to high-competitive. This differentiation occurs at younger age in female (>13 years) compared to male (>16 years) swimmers.

**Conclusion:** Pacing behavior development throughout adolescence is driven by age-related factors besides race experience. Swimmers attaining a higher performance level during adulthood exhibit a pacing behavior which better fits the task demands during adolescence. Monitoring and individually optimizing the pacing behavior of young swimmers is an important step towards elite performance.

**Key words (6/6):** Sport, race analysis, competitive swimming, future performance, talent, multilevel modelling.

## 1 1. Introduction

2 The goal-directed decision-making process regarding effort distribution (i.e., pacing) is a decisive factor for performance in exercise tasks (1, 2). The outcome of this process, the 3 4 athletes' pacing behavior, is commonly quantified by registering a measure of effort (e.g., power output or velocity) during sections of an exercise task (2, 3). Pacing seems to be learned 5 through a cyclical acquisition process, in which experience gathered during a previous task is 6 7 used to inform the athlete in future iterations of the task (4). The awareness of the benefits of distributing effort to reach a set exercise goal is first observed at 5-8 years old (5) and the 8 capability to do this effectively continues to develop during adolescence and into adulthood (6, 9 10 7). With age, the pacing behavior of children and adolescents develops to feature an increasing fit to the task demands (6, 7). Previous longitudinal studies considered the pacing behavior 11 12 exhibited by elite level adults as the endpoint of this development (6, 7). Moreover, it was revealed that athletes who reached a higher performance level in adulthood, exhibited a pacing 13 behavior resembling that of adult athletes at an earlier stage of adolescence, compared to their 14 less successful peers (6). Knowledge about the development of pacing behavior is therefore of 15 great interest for both scientists and practitioners. Unfortunately, the limited amount of 16 available research into the pacing behavior of children and adolescents consists mainly of cross-17 sectional studies with small sample sizes, often including individuals from one specific country, 18 region, school, club or team (8, 9). To provide further insights into the development of pacing 19 behavior, more rigorous longitudinal studies with large sample sizes are needed. 20

One sport in which the topic of pacing behavior has gained increasing scientific interest in the last few years, is competitive swimming (8, 10). Given the highly resistive properties of water compared to air, and the low mechanical efficiency of the swimming movement, it has been argued that adequate pacing might be more important in swimming compared to land-based sports (8, 10). Moreover, competitive swimming is a popular, global sport in which the gap

between the gold medalist and the last finisher in international competitions is decreasing (11). 26 27 In light of this, optimizing pacing behavior plays an increasingly important role in elite swimming performance (8, 10). Systematic literature reviews have shown that pacing behavior 28 of swimmers is primarily determined by the race distance and stroke type (8, 10). In races over 29 a short distance (50-100m), elite swimmers adopt an all-out pacing behavior, attempting to 30 achieve a high velocity through rapid acceleration and trying to maintain this velocity 31 throughout the race (12). During 200m races, elite swimmers adopt a fast start followed by an 32 even pace (13). Comparing different strokes, it is evident that the butterfly and breaststroke 33 events are characterized by a gradual decrease in velocity over the duration of the race, which 34 35 is mostly attributed to the relative inefficiency of these strokes compared to front crawl or backstroke. Regarding pacing behavior development in swimming, one study reported that 36 adolescent swimmers performing a 200m front crawl trial started off too fast and therefore 37 38 lacked in speed at the end of the trial (14). A second study reported that adolescent swimmers have difficulty in selecting the optimal pace, performing better in a 400m front crawl trial when 39 executing an externally imposed pace compared to a self-selected pace (15). It was proposed 40 that the difference between adolescent and adult swimmers was due to the disparity in task 41 42 experience (13, 17, 18). This, however, seems to be an oversimplification as the shift of pacing 43 behavior during adolescence is thought to originate not only from increased exercise experience but also from age-related physical maturation and cognitive development (4, 9). Additionally, 44 as the chronology of physical maturation and cognitive development processes differ between 45 46 boys and girls (19, 20), it logically follows that the timeline of pacing behavior development differs between sexes (21, 22). A profound understanding of the mechanisms behind the pacing 47 behavior of adolescent swimmers, including the influence of factors such as age, experience 48 and sex, could help coaches to guide their athletes in developing a more optimal pacing 49 behavior. 50

The present study aimed to investigate the development of pacing behavior in adolescent 51 52 swimmers, specifically disentangling the effects of age and experience and differentiating between performance levels in adulthood. It was hypothesized that the pacing behavior of 53 swimmers would develop during adolescence, gradually exhibiting more resemblance to adult 54 behavior. The demands of the task would influence the direction of the development. In short 55 tasks, the development would present itself as a change towards a more all-out pacing behavior, 56 characterized by a higher velocity during the initial stages. In longer tasks, the shift would be 57 towards a more even effort distribution. Moreover, it was hypothesized that, independent of 58 age, increased experience would facilitate a better fit with the task demands: a higher velocity 59 60 in the initial stages in the shorter tasks and an overall more even distribution of effort in longer tasks. Adolescent swimmers who eventually reached a higher performance level in adulthood 61 were hypothesized to exhibit a pacing behavior more resembling that of adult swimmers, 62 63 compared to adolescent swimmers who attained a lower performance level. As females generally exhibit puberty-related physical maturation and cognitive development at an earlier 64 age compared to their male counterparts, it was hypothesized that the split between swimmers 65 of different future performance levels would occur earlier in females compared to males. 66

67

#### 68 **2.** Methods

All procedures used in the study were approved by the Local Ethical Committee of the University Medical Center Groningen, University of Groningen, The Netherlands (201900334) in the spirit of the Helsinki Declaration. The requirement for informed consent of the participants was waived given the fact that the study involved the analysis of publicly available data and analyses were group-based.

74

#### 75 2.1. Data collection

All available 100m and 200m freestyle long course performance data (i.e., date of the race, total race time and available 50m split times) of both male and female swimmers performing between 2000 and 2021, were collected from Swimrankings' database (<u>www.swimrankings.net</u>). This resulted in 2,857,181 (100m freestyle) and 1,897,872 (200m freestyle) observations. The assumption was made that all swimmers chose the front crawl during the freestyle events. Performance data were collected from 113 countries across the world. The date of birth of all included swimmers was collected using the same database.

83

# 84 2.2. Data processing

Swim performances over 180s (100m freestyle) and 360s (200m freestyle) were excluded from 85 the analysis to ensure a homogeneous dataset. Performance data were classified per swimming 86 season, starting on the 1<sup>st</sup> of September and ending on the 31st of August of the next calendar 87 88 year. Data from the 1st of January 2008-2010 were excluded from analysis, because of the impact of full-body polyurethane swimsuits on swimming performance in that period (23-25). 89 Performance data from season 2019-2020 were excluded as competitions and training 90 opportunities were disturbed because of the COVID-19 pandemic. A total of 2,773,387 91 92 observations (100m freestyle) and 1,842,992 (200m freestyle) observations remained. For each 93 swimmer, the Season Best Time (SBT) per swimming season was used for further analysis. Age at SBT was determined using the swimmer's date of birth. Race experience was defined 94 as the cumulative number of races of a specific event, which the swimmers had completed 95 before SBT. 96

97

# 98 2.3. Inclusion criteria

99 For the purpose of this study, it was important to outline the development of pacing behavior100 from a young age on toward the age of peak performance. Peak performance in competitive

swimming is on reached at 24 ( $\pm$ 2) years for males and at 22 ( $\pm$ 2) years for females (26). 101 102 Therefore, only swimmers who had at least one swim performance in the age category of 22 years or older (male) or 20 years or older (female) were included. To ensure a dataset 103 representing the developmental pathway of pacing behavior towards peak performance, swim 104 performances after the swimmer's career-best swim performance were excluded. To 105 longitudinally study pacing behavior development, included swimmers had to be between 12 106 107 and 24 years old and have performance data with 50m split times in at least three swimming seasons. To study pacing behavior independent of current performance, split times of each 50m 108 section were converted into relative section times (RST), representing the percentage of the 109 110 total race time spent in one section. The inclusion criteria were conducted for the 100m and 111 200m events separately.

112

Swim performances of multiple generations (i.e., from 2000 through 2021) were included in 113 the dataset, which necessitated the correction of evolution in competitive swimming. As such, 114 swim performances were defined as a percentage of the prevailing world record (WR) of the 115 corresponding sex, referred to as relative Season Best Time (rSBT) (27, 28). World records 116 from 2008 and 2009 were replaced by the prevailing fastest time in a textile swimsuit. 117 118 According to the event, swimmers were allocated to the elite, sub-elite or high-competitive performance group by using their event-specific all-time rSBT after 20 (female) or 22 (male) 119 years of age (see Table 1). The elite level was defined as the average rSBT of the 50<sup>th</sup> swimmer 120 of the event-specific FINA World Ranking List between 2016 and 2021 (11). Sub-elite level 121 and high-competitive level were defined as the average rSBT of the 8<sup>th</sup> and 50<sup>th</sup> swimmer of the 122 event-specific National Ranking List of the Netherlands between 2016 and 2021 (11). 123 Swimmers with a best rSBT outside the limits of the high-competitive group were excluded 124 from further analysis. For the 100m event, this resulted in 3,498 swimmers (1,659 female) with 125

126 15,960 observations (7,384 female) with an average of  $5.3 \pm 1.9$  observations per swimmer. For 127 the 200m event, this resulted in 2,230 swimmers (1,252 female) with 10,309 observations 128 (5,412 female) with an average of  $5.3 \pm 1.9$  observations per swimmer.

129

130 2.4. Statistical analysis

Following the methods introduced by Menting et al. (7), longitudinal multilevel models were 131 created to describe pacing behavior as a function of age, race experience and performance 132 group. Multilevel modelling allows for the creation of models in which repeated measures (level 133 1) are nested within individual swimmers (level 2), allowing the use of longitudinal data with 134 135 varying number of measurements between swimmers as well as a variety in temporal spacing between measurements. Analyses were performed using the lmer4 package in R (R version 136 3.6.0)(29, 30). Statistical assumptions (e.g., multicollinearity) were checked and outliers were 137 screened and removed (100m: 915, 200m: 1,006). The RST per 50m section were included as 138 dependent variables. In contrast to split times, all RST must add up to 100%. With respect to 139 this constraint, one out of two (100m freestyle) and three out of four (200m freestyle) multilevel 140 models were created. The remaining, free section (RST 50-100m in both events) was calculated 141 142 from these models. Following that the sum of 50m sections must add up to 100%, the same 143 predictor variables (fixed part) and variance structure (random part) had to be incorporated into each model equation. Predictor variables age and race experience were included as continuous, 144 time-varying factors whereas performance group was included as a categorical, time-invariant 145 146 factor. The power law of practice states that the effect of experience on performance decreases as the level of experience increases (31). In addition, the age effect on performance decreases 147 as swimmers are fully matured (27). As such, the effect of a 1-year increase at age 13 will be 148 larger than a 1-year increase at age 19. To account for this, the variables age and race experience 149 were log-transformed, of which the latter transformation was needed to meet the assumption of 150

normality. To represent the three performance groups in the statistical models, two dummy variables (sub-elite and high-competitive) were included and the elite group functioned as reference level. A random intercept model was selected as the most appropriate variance structure, allowing the inclusion of each swimmer's individual trajectory that randomly deviates from the average population trajectory. In sum, the following multilevel model was adopted:

157

158  $RST_{is} = \alpha + \beta_1 \times \log (Age_{is}) + \beta_2 \times \log (RaceExperience_{is}) + \beta_3 \times SubElite_i$ 

159 +  $\beta_4 \times HighCompetitive_i + u_i + \varepsilon_{is}$ 

160 
$$u_i \sim N(0, \sigma_0^2)$$

- 161  $\varepsilon_{is} \sim N(0, \sigma^2)$
- 162

 $RST_{is}$  was the relative split time of a 50m section for swimming season s of swimmer i,  $\alpha$  the 163 intercept assigned to the elite group,  $Age_{is}$  the corresponding age value,  $RaceExperience_{is}$ , 164 the corresponding race experience value,  $SubElite_i$  the dummy variable of swimmer *i* assigned 165 to the sub-elite group and  $HighCompetitive_i$  the dummy variable of swimmer *i* assigned to 166 the high-competitive group. The unexplained information was the sum of  $u_i$  (between-subject 167 variance) and  $\varepsilon_{is}$  (residual variance). The models were validated by using graphical tools to 168 check violations of homogeneity, normality and independence. Predictor variables were 169 considered significant if the estimated coefficient is greater than twice the standard error of the 170 estimate (p<0.05). Post-hoc analyses were performed for models with future performance group 171 172 as significant predictor variable. For this analysis, swimmers were classified in age categories based on their age on the 31st of December of the swimming season. Per age category, an 173 independent sample t-test was conducted to examine from which age onward between-group 174 differences in pacing behavior occurred. These follow-up analyses were executed for age 175

176 categories with at least 30 observations per performance group. For all tests, p < 0.05 (two-177 tailed) was set as significance.

178

# 179 **3. Results**

The models created can be found in Table 2. Using the fixed part of the models, predictions for the dependent variables can be made. For example, for the RST in the 100-150m segment of a 200m event performed by an 18-year-old male swimmer, with 20 previous races and an adult performance level as high-competitive, the following value will be predicted as:

184

185  $RST \ 150m = 27.42 + (-0.55 \times \log 18) + (-0.03 \times \log 20) + (-0.00 \times 0) + (0.09 \times 1)$ 186 = 25.83%

187

188 3.1. Age

The predicted effect of age on RST is visualized in Figure 1A (100m) and Figure 2A (200m). Older male swimmers were relatively faster in the first 50m of the 100m. No effect of age was indicated in female 100m swimmers. In the 200m, older male and female swimmers were predicted to start relatively slower, have a relatively faster middle section and a relatively slower final 50m section compared with their younger counterparts.

194

# 195 3.2. <u>Race experience</u>

196 Race experience significantly impacted RST in all segments except for the final segment in the 197 male 200m event, as visualized in Figure 1B (100m) and Figure 2B (200m). In the 100m, more 198 experienced male and female swimmers were relatively slower in the first half of the race. In 199 the 200m, male swimmers with more race experience were relatively slower in the first 50m

10

section, but faster in the 150m section. More experienced female swimmers were relatively
slower in the first 50m section and relatively faster in the 150m and 200m sections.

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# 203 3.3. <u>Performance level</u>

Elite male swimmers were faster in the first 50m of the 100m, compared to the high-competitive 204 group. Post hoc analysis revealed that the male swimmers of the elite group started 205 206 differentiating themselves at 17 years old ( $t_{(99.6)} = -2.21$ , p<0.05). No difference was found between female swimmers of differing performance groups. In the 200m, elite male swimmers 207 were predicted to be relatively slower in the first 50m, but faster in the 150m section, compared 208 209 to swimmers from the high-competitive group. Swimmers from the elite group differentiated themselves as early as 16 years old (RST50: t(51.728)= 3.10, p < 0.01; RST150: t(57.699) = 3.11, p 210 < 0.01). Elite female swimmers were relatively slower in the first 50m section, but faster in the 211 212 150m and 200m sections, compared to the high-competitive group. The difference started at 13 years of age (RST50:  $t_{(51.07)} = 2.36$ , p < 0.05, RST150:  $t_{(77,62)} = 4.62$ , p < 0.001; RST200:  $t_{(97,66)}$ 213 = -3.065, p < 0.01). In both the 100m and 200m, the model predicted no significant difference 214 in RST between the elite and sub-elite groups (Figure 1C and Figure 2C). 215

216

# 217 **4. Discussion**

The present study aimed to investigate the pacing behavior development of swimmers throughout adolescence, explicitly differentiating between the effects of age and experience as well as investigating its relationship to performance level in adulthood. As hypothesized, older male swimmers adopted a more all-out distribution of effort in the 100m event, although this development was not exhibited by female swimmers. In the 200m, male and female swimmers exhibited a more even distribution of effort as they became older. Both race experience and age independently impacted the pacing behavior of adolescent swimmers, providing evidence that experience is not the sole driver of pacing behavior development. Furthermore, adolescent swimmers who in adulthood reached the elite level (100m: male, 200m: male & female) exhibited a pacing behavior more resembling adult swimmers compared to swimmers in the high-competitive group. As hypothesized, the distinction in pacing behavior between swimmers of differing future performance level occurred earlier in female compared to male swimmers.

230

231 *4.1*.

# 4.1. Pacing behavior development in swimming

232 In previous literature, the effect of experience and age has often been used synonymously (13, 17, 18). However, this seems to be an oversimplification. In the 100m, the behavior of older 233 male swimmers moves towards a fast first 50m, hereby paralleling the behavior of the elite 234 swimmers in adulthood. This resemblance, however, was not observed when comparing male 235 swimmers based on race experience. It supports the notion that pacing behavior development 236 237 is driven by other age-related factors (e.g., physical maturation and cognitive development) alongside the increase in experience. Additionally, these findings suggest that race experience 238 in itself may not be sufficient to explain the development of future elite performers. Further 239 240 evidence for this view is provided by the finding that in the 200m event, age still impacts on pacing behavior in both male and female swimmers, even with a separate variable for race 241 experience included in the model. Moreover, the results show that in line with the hypothesis, 242 the separation between future performance levels occurs at a younger age in females (13 years 243 244 old) compared to males (16 years old). The earlier onset of pacing behavior development in 245 females which has previously been described in a cross-sectional study (21) is thereby confirmed by the current longitudinal study and is thought to be caused by the earlier onset of 246 physical maturation and cognitive development (21, 22). 247

Based on previous literature, it was proposed that with experience and age, adolescent athletesadapt their pacing behavior to better fit the task demands (6, 7). Indeed, within the present study,

12

there is a difference in the development of pacing behavior in the 100m and the 200m events. 250 251 In the 100m event, older male swimmers adopt a more all-out pacing behavior, characterized by a relatively faster first lap. The relatively faster initial 50m could be the result of an improved 252 race start, including the dive and underwater phase. Alternatively, it has been established that 253 in tasks of similar duration to the 100m freestyle event, better-performing athletes differentiated 254 themselves by a relatively more all-out pacing behavior (32, 33). De Koning et al proposed that 255 256 for shorter events (<2min), the advantage of a higher velocity in the first part of an exercise task and the lower amount of kinetic energy left at the end of the race, outweighed the disadvantage 257 of higher frictional losses associated with the higher average velocity (33), which was further 258 259 evidenced through modelling studies in speed skating and track cycling (34, 35), though 260 differences between sports were visible (36). Indeed, elite swimmers competing in the 100m freestyle finals of international events exhibited an all-out pacing behavior, comparable to the 261 262 one found in the current study (12). Moreover, it was reported that elite male swimmers adopted a more all-out pacing behavior (RST50m: 47.91%, RST100m: 52.09%) compared to female 263 swimmers (RST 0m: 48.29%, RST100m: 51.77%) (12). These findings are supported by the 264 results of the present study, as adolescent male swimmers not only presented a more all-out 265 266 pacing behavior, but also continued to develop this behavior with age. The reason behind the 267 apparent difference in pacing behavior between male and female swimmers could potentially 268 be found in the physical and physiological differences between male and female swimmers (37). Alternatively, it has been reported that males engage more in risk-taking behavior and 269 270 therefore are expected to generally adopt a more all-out pacing behavior (38).

Contrary to the 100m event, older male and female swimmers adopt a relatively more even
distribution of velocity in the 200m event. This is achieved by a relatively slower first and last
50m section and a relatively faster middle section. Swimming is a head-to-head type event, as
the winner of a race is the swimmer who covers the given distance before the other swimmers,

independent of the time set by swimmers in previous races (8). Remarkably, the development 275 of pacing behavior in swimming does not resemble that of other middle-distance head-to-head 276 events, such as short-track speed skating. Studies in these events have reported that the athletes' 277 pacing behavior develops towards a more conservative start and middle section of the race to 278 facilitate the athlete to position themselves well and be relatively faster in the key final stages 279 of the race (7, 21, 22). The development of pacing behavior in the 200m more resembles the 280 one found in time-trials of a similar duration (6, 39, 40). This development is characterized by 281 a shift towards a more even distribution of effort, which allows for a minimization of energy 282 loss due to acceleration and deceleration, resulting in better performance in middle- and long-283 284 distance time-trial based events (41). This resemblance to time-trials likely originates from the 285 lane-based nature of competitive swimming (8). The lanes inhibit the interaction with other competitors, resulting in a less interactive competitive environment as is also found in time-286 287 trial events. Taken together, coaches could expect to encounter sex- and age-related differences in pacing behavior in adolescent swimmers of the same level of race experience. Additionally, 288 as adolescent athletes get older, they adapt their pacing behavior to fit the characteristics of the 289 task, with male swimmers adopting a more all-out behavior on the 100m and both male and 290 291 female swimmers adopting a more even distribution of effort in the 200m event.

292

#### 293 4.2. Future performance

The findings of the present study provide evidence that the swimmers who perform within 104% of the prevailing world record as adults (i.e., the elite group), exhibit pacing behavior that differentiates them from other adolescent swimmers (i.e., the high-competitive group). It therefore establishes that adequate pacing behavior development is an essential part of the developmental pathway towards elite swimming performance. In the 200m event, the effect of future performance level parallels the effects of age and race experience in both males and

females. In other words, swimmers that achieve a higher level of performance in adulthood, 300 exhibited a pacing behavior resembling that of older and more experienced swimmers during 301 adolescence. This is different for the 100m event. Adolescent male swimmers who reach the 302 elite level as an adult, exhibit a pacing behavior that is more resembling the pacing behavior of 303 the older swimmers (all-out pacing behavior) compared to that of their peers who reach the 304 high-competitive level. However, the current findings suggest that more race experience results 305 in a more conservative first 50m in the 100m instead of going more all-out. The underlying 306 mechanism for this converse effect of race experience on pacing behavior in 100m event 307 remains unclear and warrants further research. In females no effect of either performance level 308 309 or age was found, however the effect of race experience was equal to males.

310 In the present study, no distinction could be made between elite and sub-elite swimmers. A 311 possible reason for this could be the high performance level of all included swimmers in the present study. To place it into context, for a male 200m swimmer competing in 2022, the 312 performance levels equal a time of <106.18s (elite), 106.18-109.75s (sub-elite) and 109.75-313 118.93 (high-competitive). The Olympic Qualifying Time for Tokyo 2021 was set at 107.02s 314 (42). In comparison to the current study, a previous study did report a difference in pacing 315 behavior between three performance levels (6). However, Wiersma et al determined adult 316 317 performance using the season best performance at 18-19 years of age, whereas the present study 318 used a more appropriate measure to indicate adult performance level: all-time peak performance after 20 (female) or 22 (male) years of age expressed as a percentage of the prevailing world 319 record. Recalculating the performance level of the athletes in the previous study, using these 320 321 methods results in a much wider spectrum of performance (elite: 113.8%, sub-elite: 120.6%, non-elite: 129.7%), could explain why the previous study did find a difference in pacing 322 behavior development between the performance levels. 323

324

#### 325 4.3. Limitations and future directions

Although the models created in the present study provide novel insights into the relationship 326 327 between age, experience and pacing behavior, the models do not account for all the variance in a swimmers' pacing behavior. Pacing is a complex, psychophysiological process and even when 328 the task characteristics are set, it is influenced by a multitude of factors relating to the individual 329 (i.e., physical maturity, cognitive development, muscle fiber type distribution) and environment 330 (i.e., coaching culture, training opportunities) (1, 9, 43, 44). The absence of these factors has 331 332 potentially led to the lower explained variance of the models. For example, there was no effect for age or performance level on pacing behavior in female swimmers competing in the 100m 333 334 event. In males, the effect of age and performance group was also more pronounced in the 200m 335 event compared to the 100m event. It could be that 100m freestyle performance is 336 predominantly driven by the development of physical characteristics, such as muscle fiber type distribution, whereas in the 200m event the distribution of effort is a larger determination factor 337 338 in the outcome of the race. However, another reason might be that the 100m freestyle is often contested by both 50m and 200m specialists. The energetic system requirements between the 339 340 50m and 200m freestyle events differ significantly and therefore swimmers who compete in these events are adapted to physiologically very different tasks (37), therefore exhibiting a 341 342 different pacing behavior. The coming together of these two types of specialized swimmers 343 might have impacted the results of the present study. It should be pointed out that previous studies have evidenced that swimming performance is impacted by velocity in free swimming 344 sections, but also by turns and underwater phases (45). Quantification using 25m or even 5m 345 346 and 10m sections has previously been demonstrated to reveal more detailed definitions of impact of these factors on a swimmers' performance (18, 45). However, these data have to be 347 gathered using camera set-ups and specialized software, which drastically decreases practicality 348 and would have reduced the sample size greatly. In the end, the present study aimed to create 349

models which could provide insight into the relation between age, experience and future 350 351 performance level, not precisely predict each individual swimmers' pacing behavior. The large sample size, consisting of swimmers from five continents, and the strong longitudinal nature of 352 the data are of key importance to the rigidity of the present study's design, not in the first place 353 because more large scale longitudinal studies on pacing behavior development are needed (4, 354 22). Consequently, the decision was made to use publicly available 50m split times. The choice 355 for this approach does allow for future studies, using more detailed quantifications of pacing 356 behavior and the inclusion of more individual and environmental factors, to provide additional 357 insights into the development of pacing behavior in the 100m and 200m freestyle events. 358

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## 360 *4.4. Practical application*

The effect of age and race experience on pacing behavior as reported in the present study are 361 362 relatively small compared to that of task defining characteristics such as race duration or stroke type (8). However, in a 200m freestyle, an average 0.16% difference in velocity distribution per 363 50m section (the difference between a 12 and 18-year-old male swimmer as calculated using 364 365 the models in the present study) constitutes 0.20s. In a sport where 0.01 of a second can be the difference between winning and losing, a 0.20s difference in velocity distribution in every 50m 366 section can indeed have a very real impact on competition performance. Using the formula 367 provided in the present study, coaches could determine whether their swimmers are on track of 368 369 developing the pacing behavior necessary to achieve the elite performance level. One point of 370 notice should be made to this approach: the road to elite performance is not always linear and pacing is only a part of the skillset necessary to reach the top (46). In addition, it has been 371 established that to pace adequately, athletes need to match their personal performance capacities 372 to the task demands. Seeing as there is variation in each swimmer's performance capacities, a 373 slightly different pacing behavior could be optimal for each swimmer. It is therefore important 374

375 to take the outcomes of the formula from the present study as a starting point and take an individualized approach to the development of each swimmer. Within this approach, coaches 376 are advised to provide the swimmers with opportunities to experiment with variants of their 377 established pacing behavior (4). Introducing variability would provide swimmers with the 378 opportunity to discover a more optimal match between their personal performance capacities 379 and the task demands (47). Coaches could induce this variation by providing augmented 380 feedback via tools such as a stopwatch, pacer clock, wearable metronome, underwater lights or 381 smart goggles (48). Demonstrating this method, a recent study reported that a three week 382 training program in which adolescent swimmers were provided with feedback on their own 383 384 pacing behavior was effective in increasing 400m freestyle performance (49). Subsequently, 385 practice of the new variation of pacing behavior could be further increased by gradually taking away sources of feedback and adding environmental factors such as opponents, therefore 386 387 training the swimmers to maintain their capability of decision-making regarding effort distribution in a more realistic competitive environment (22, 48). 388

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#### 390 5. Conclusion

391 The current large-scale study is the first in its kind in that it investigates the pacing behavior of swimmers from five continents over a period spanning the last twenty years. The rigorous 392 multilevel modelling approach with corrections for prevailing world records revealed insights 393 394 on developmental patterns based on thousands of swimmers with on average five competitive seasons in adolescence. The pacing behavior of swimmers develops during adolescence, as 395 396 older swimmers adopt a pacing behavior that better suits the task demands (100m: more all-out 397 [males only], 200m: more even). Although swimming is a head-to-head type of competition, the development of pacing behavior resembles that of time-trial events, most likely due to the 398 lane-based nature of the sport. The persistence of the effect of age on pacing behavior when 399

race experience was also included as predicting variable, supports the hypothesis that pacing 400 401 behavior development during adolescence is driven by other factors in addition to increased experience, such as physical maturation and cognitive development. Swimmers who reach the 402 elite performance level in adulthood, exhibit a pacing behavior better suits the task demands 403 and that resembles that of adults (100m: more all-out [only males], 200m: more even) during 404 adolescence. In the 200m, this differentiation occurs earlier in females compared to males, most 405 406 likely due to the earlier onset of age-related physical maturation and cognitive development in females. Coaches are advised to take notice of the complex development of pacing behavior 407 which occurs throughout adolescence. Furthermore, coaches could use the data presented in the 408 409 present study as a starting point for an individualized approach to optimize the pacing behavior development in their swimmers and better guide them on the road towards elite performance. 410

#### 411 **6.** Acknowledgments

#### 412 6.1. Authors' contributions

The study conception and design were done in full collaboration with all authors. SGPM and
AKP contributed equally to this manuscript. All authors critically revised the work. All authors
read and approved the final manuscript.

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## 417 6.2. Conflict of interest

The authors do not have any conflict of interest. The authors declare that the results of the study are presented clearly, honestly, and without fabrication, falsification, or inappropriate data manipulation. The results of the present study do not constitute endorsement by the American College of Sports Medicine. The authors received no specific funding for this work.

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# 423 **7. Tables and figures**

Table 1. The total number of swimmers and observations according to sex, performance leveland event included in the analysis.

Table 2. Multilevel models predicting relative section time per 50m section, divided by sexand event.

428 Figure 1. Predicted pacing behavior for males and females in the 100m freestyle event429 according to age, race experience and performance level.

430 Figure 2. Predicted pacing behavior for males and females in the 200m freestyle event431 according to age, race experience and performance level.

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