

1 **Pacing profiles, variability and progression in 400, 800, and 1500-m freestyle**
2 **swimming events at the 2021 European Championship**

3

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25 **swimming events at the 2021 European Championship**

26

27 **ABSTRACT**

28 Performance intra-variability has not been studied in long-distance swimming. The aims
29 were to describe the pacing profile in 400-, 800- and 1500-m freestyle events and to
30 analyse the variability (coefficient of variation[CV]) and performance progression (% Δ)
31 within and between rounds. A total of 256 swims of 130 elite-swimmers (70 males and
32 60 females) were analysed at the 2021 European-Championship (indoor long-course).
33 Linear mixed-effect models were applied for each swimmer and race-time performances
34 to obtain the CV and % Δ between each lap and rounds (i.e. heats and final). T-test was
35 conducted to compare the CV between medallist and non-medallist. First and last laps
36 were the fastest ($p < 0.001$) in all events compared to the intermediate laps which showed
37 an evenly-pace. Parabolic pacing profile was adopted in all events. Male swimmers
38 obtained a CV-average of $0.52 \pm 0.49\%$ between rounds ($-0.64 \pm 0.8\%\Delta$) and females, a
39 CV-average of 0.70 ± 0.45 ($-0.71 \pm 0.92\%\Delta$). Medallist swimmers obtained higher CV
40 between rounds (1.00–1.08%) compared to non-medallist finalist (0.22–0.47%).
41 Parabolic pacing profiles were adopted in 400-, 800-, and 1500-m races. The best
42 swimmers adopt conservative strategies in heats to improve their performance in final,
43 obtaining higher CV and % Δ between rounds.

44

45 **Keywords:** competition analysis, elite swimmers, performance, strategies, variation.

46

47 INTRODUCTION

48

49 In endurance performance, athletes should manage their energy expenditure during a
50 competition (Foster et al., 2012). For sport scientists, this phenomenon is known as pacing
51 or pacing strategy (Abbiss & Laursen, 2008). The optimal energy distribution depends on
52 each athlete and their own metabolic characteristics, although it can also be affected by
53 external factors that influence pacing (e.g., environmental elements) (Roelands et al.,
54 2013). Therefore, athletes might develop different pre-planned pacing strategies to
55 optimise their performance during the competition (Edwards & Polman, 2013). Based on
56 pacing strategy or how the athlete distribute their energy expenditure, some pacing
57 profiles have been identified in endurance sports, like positive (i.e., performance
58 decrease), negative (i.e., performance increase), or even (i.e., evenly performance)
59 (Abbiss & Laursen, 2008).

60

61 Pacing strategies are more important in swimming than in other cyclic sports due to the
62 effect of the resistance forces (e.g., hydrodynamic drag), which increases exponentially
63 with an increase in velocity (Marinho et al., 2011); hence, velocity variations during the
64 race could lead to higher energy cost for swimmers (Mauger et al., 2012). Pacing analysis
65 have been considered useful in middle-distance events like 400-m and, especially, in
66 long-distance events like 800- and 1500-m freestyle because of the large number of laps
67 and times performed by swimmers (Lipińska et al., 2016b). Previous studies have
68 described the pacing profile in the referred events (i.e., 400-, 800-, and 1500-m), where it
69 has been identified a parabolic or U-shaped profile (Lipińska et al., 2016a; Lipińska et
70 al., 2016b; Mauger et al., 2012). Those profiles are characterized by a fast start followed

71 by an evenly paced mid-section and a fast end-spurt in the later laps of the race (Skorski
72 et al., 2014). Moreover, in 800- and 1500-m events, elite swimmers adopt a small decrease
73 in performance in the intermediate laps (McGibbon et al., 2018). Therefore, this suggests
74 that swimmers may employ different tactics or strategies to conserve energy during the
75 race.

76

77 The analysis of variability and progressions in performance times within and between
78 races is essential in swimming competitions (Stewart & Hopkins, 2000). The coefficient
79 of variation (CV) is used for the study of performance variability, which is defined as the
80 percentage of random variation in the athlete performance (Hopkins et al., 1999). In major
81 swimming events, like European, World Championships, or Olympic Games, swimmers
82 can participate in several races with short recovery periods between heats, semi-finals,
83 and finals. Hence, the swimmers' progression must guarantee the qualification for the
84 semi-final and later for the final, where the peak performance must be achieved (Mujika
85 et al., 2019; Pyne et al., 2004). In addition, to reduce the accumulated fatigue during the
86 previous round and to reach the performance peak at the final, swimmers should adopt
87 pacing patterns to reduce as much as possible the energy expenditure (Foster et al., 2003;
88 Mauger et al., 2012).

89

90 In a major competition, like the 2021 European Championship, swimmers' performance
91 progression between rounds is required to get a medal (Mujika et al., 2019). In contrast
92 with other swimming events (e.g., 50-, 100-, or 200-m) where heats, semi-finals, and
93 finals are performed, in 400-, 800-, and 1500-m freestyle races only two rounds are
94 performed (i.e., heats and final). Thus, the best swimmers should improve by ~1% their

95 performance variability between heats and final to increase the chances of success (Pyne
96 et al., 2004). To appreciate this difference between swimmers' level, an interesting option
97 might be to study the CV values between rounds in medallist and non-medallist
98 swimmers. Hence, performance variability and the values obtained in CV within and
99 between rounds must be considered to get relevant information about the current
100 competitive swimming and thus, the swimmers' progression.

101

102 Knowing the relevance of pacing and performance variability, there are no recent studies
103 that bring together the pacing profiles, CV, and performance progressions in 400-, 800-,
104 and 1500-m in major swimming events. Therefore, the aims of the present study were (1)
105 to describe the pacing profile in 400-, 800-, and 1500-m freestyle events and, (2) to
106 analyse the variability (coefficient of variation [CV]), and performance progression (% Δ)
107 within and between rounds. It was hypothesized that swimmers would adopt a parabolic
108 profile in the three events. The first and last laps would be the fastest laps, while the
109 intermediate laps would present an evenly pace. The best swimmers (i.e., medallist)
110 would adopt a conservative pacing strategy in the heats, presenting a higher CV values
111 and progression than non-medallist in the final.

112

113 **MATERIALS AND METHODS**

114

115 **Subjects**

116 A total of 256 swims (208 heats and 48 finals) of 130 elite swimmers (70 males [age:
117 22.21 ± 3.22 years] and 60 females [age: 20.68 ± 3.64 years]) were analysed, being 108
118 of 400-m races, 85 of 800-m races, and 63 of 1500-m races.

119

120 **Data collection**

121 Official race time and 50-m lap times for 400-, 800- and 1500-m of elite swimmers at the
122 2021 European Championships (indoor long-course) were obtained from the official web
123 site of the European Swimming League: (www.len.eu). Informed consent and ethical
124 approval were not required, since all the information is available in the public domain.

125

126 For each event, the results of the two rounds (i.e., heats and final) were analysed. The
127 official data was downloaded by applying a Web Scraping routine in Python®. Once the
128 automated process was completed, two independent researchers checked that the results
129 of all events had been downloaded and verified that no information was missing. The
130 downloaded data consisted of "distance", "stroke", "round", "rank", "lane", "swimmer
131 name", "reaction time", "lap times", and "final time". Subsequently, the following
132 variables were calculated:

133

- 134 - The pacing laps CV, which represents the pacing lap time variability (i.e., every
135 50-m) in each event between rounds (i.e., heats and final).
- 136 - The intra-athlete CV, which represents the random variation in performance
137 between rounds, obtained between heats and final (Hopkins et al., 1999).
- 138 - The inter-athlete CV, which represents the dispersion of ability among athletes in
139 the two rounds. Two different inter-athlete CVs were obtained: 1) obtained from

140 the performance of the participants in the heats; and 2) obtained from the
141 performance of the finalists.

142 - Relative change (% Δ) in performance between rounds was obtained with the
143 following formula:

$$\% \Delta = \frac{\text{Round 2 performance} - \text{Round 1 performance}}{\text{Round 1 performance}} \times 100$$

144

145 where Round 2 performance refers to the race time of the final and Round 1
146 performance refers to the race times of the heats. The criterion for performance
147 progression, no change, or regression was % Δ being lower, equal, or higher than
148 0, respectively (Mujika et al., 2019).

149 - Relative change (% Δ) in laps performance between rounds was calculated by the
150 average of the % Δ between the heat and the final for the finalists in every lap.

151

152 **Statistical Analysis**

153 Statistical procedures were carried out using SPSS 24.0 (IBM, Chicago, IL, USA). The
154 normality of the distribution was inspected with Kolmogorov Smirnov test and the
155 homoscedasticity was verified with the Levene test. All analysis were conducted
156 differentially by distance and sex (Shapiro et al., 2021). Average times for 50-m laps were
157 obtained for the pacing profiles analysis. Linear mixed-effects models were applied in
158 finalist swimmers to obtain CV and changes for each lap and also it was applied for all
159 swimmers and race time performances to obtain the CV and % Δ between rounds (i.e.,
160 heats and final). Independent samples t-test were conducted to compare the mean
161 difference of CV between medallist and non-medallist swimmers. Pacing variability was

162 assessed through repeated-measures analysis of variance (ANOVA) to analyse the
163 variation per 50-m lap and Bonferroni post-hoc test was used to verify significant
164 differences between each pairwise. The same test was applied to explore differences in
165 CV and % Δ between distances. Significance level was set at $p < 0.05$ for all the analysis.

166

167 **RESULTS**

168

169 The pacing profiles for finalist swimmers in heats and finals in 400-, 800-, and 1500-m
170 are shown in Figures 1, 2, and 3, respectively. The highest values of the pacing laps CV
171 in all events were obtained in the last lap, except in the 1500-m females where it was
172 obtained in the second lap. The first lap was significantly faster ($p < 0.001$) than the rest
173 of laps in the 400- and 800-m races in both sexes, and in 1500-m females. In 400-m and
174 800-m races across both sexes, the last lap was faster ($p < 0.001$) than the rest, except
175 with the second lap where no significant differences were found ($p > 0.05$). In the case of
176 1500-m males, first and last lap were faster than the rest ($p < 0.001$), but no differences
177 were found between them ($p > 0.05$). In 1500-m females, last lap was significantly faster
178 ($p < 0.05$) than the first and laps from 15 to 29.

179

180 (Please insert Figure 1 near here)

181

182 (Please insert Figure 2 near here)

183

184 (Please insert Figure 3 near here)

185

186 The results of the linear mixed-effects model analysis, within-subject CVs and % Δ
187 between the two rounds and distances, are presented in Table 1. Inter-athlete CVs in heats
188 and finals, are presented in Table 2. The linear mixed-effects model analysis revealed
189 interactions in CV and % Δ for swimmers who qualified for finals compared to heats, with
190 62% and 50% of the female and male swimmers, respectively, obtaining a CV greater
191 than 0.4% and with 83% of both female and male swimmers achieving performance
192 improvements. Independent samples t-test showed higher CVs and % Δ for medallist than
193 non-medallist swimmers (Table 3). One-way ANOVA testing revealed no differences in
194 within-subject CV and % Δ between the heats and finals ($p > 0.05$).

195

196 (Please insert Table 1 near here)

197

198 (Please insert Table 2 near here)

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200 (Please insert Table 3 near here)

201

202 **DISCUSSION**

203

204 The aims of the present study were to describe the pacing profile, and to analyse the
205 variability and performance progression in 400-, 800-, and 1500-m elite swimmers. The
206 first and last laps were the fastest laps in all events. In the intermediate laps, swimmers
207 adopted an evenly pace. The parabolic pacing profile was adopted in 400-, 800-, and

208 1500-m races. In the three events analysed, male swimmers obtained a CV average of
209 0.52% between rounds and females a CV average of 0.70%. As it was hypothesized,
210 medallist swimmers obtained higher CV values between rounds compared to non-
211 medallist.

212

213 Parabolic pacing profiles adopted by the participants of the European Championship in
214 400-, 800-, and 1500-m (Figures 1, 2, and 3) are in agreement with the profiles observed
215 in previous studies (Lipińska et al., 2016a, 2016b; Mauger et al., 2012). Pacing in
216 swimming competitions is highly influenced by the start in the first lap and by an
217 emphasis to finish the race in the last lap (Lipińska et al., 2016b), as observed in the
218 present study. The faster lap times in the first one are a consequence of the dive and
219 underwater undulatory swimming, thus, the second lap was possibly also faster by the
220 contribution of the faster swimming speed of the first lap (Lipińska et al., 2016b).
221 Furthermore, pacing change in the last two laps of 800- and 1500-m events has also been
222 specifically analysed (Neuloh et al., 2020). The end-spurt was evident in both swimming
223 events and also in 400-m, where swimmers increased their swimming velocity in the last
224 lap. Besides, as it was observed the end-spurt in 800- and 1500-m events was crucial to
225 achieve a medal, being a more pronounced end-spurt in medallist compared to non-
226 medallists swimmers (Neuloh et al., 2020). Hence, the pacing behaviour during the
227 competition, like the cited end-spurt, could explain the highest values of the pacing laps
228 CV obtained in the last laps.

229

230 At the 2021 European Championship, a large number of swimmers performed 800- and
231 1500-m events. Six of the eight female finalist swimmers repeated in 800- and 1500-m

232 events, even two of these females swam also the 400-m final. In the case of males, three
233 of the eight swimmers repeated in 800- and 1500-m finals. The similarity of energy
234 requirements between these swimming events causes swimmers to double their efforts in
235 major championships (Pyne & Sharp, 2014). Likewise, swimmers might develop their
236 own performance templates (i.e., pacing strategy) based on their experience, to improve
237 their performance in the different rounds, or even, to impose their preferred pace from the
238 start of the event (Foster et al., 2009). These reasons could justify the similar pacing
239 strategies between the 800- and 1500-m, since high percentages of swimmers repeated
240 events. Hence, pacing could have a significant role in these events as all swimmers chose
241 the same strategy, regardless of their final result or their sex (Lara & Del Coso, 2021).

242

243 Although it has previously shown that at least a 0.5% progression in swimmers'
244 performance is needed to be considered an effective strategy for success (Stewart &
245 Hopkins, 2000), several studies have reported different values within and between
246 swimming races (Pyne et al., 2004). Similar performance improvements on variability
247 compared with the results of the present study has been reported between performance
248 races for junior and elite swimmers, where it was observed an improvement of 1.2%
249 between heats and final (Skorski et al., 2014; Skorski et al., 2013). A study with
250 Australian and US Olympic sprint distances swimmers (i.e., 50-, 100-, and 200-m events),
251 it was observed a performance improvement of 0.6-0.7% and 0.5-0.7% respectively
252 between semi-finals and finals (Pyne et al., 2004). On the other hand, for the analysed
253 400-, 800-, and 1500-m races, higher values were described than those obtained at the
254 2021 European Championship (Tables 1 and 2), reporting a 1.1% for 400-m and 1.4% for
255 800- and 1500-m of average performance improvements between heats and finals (Pyne
256 et al., 2004). The within swimmer CVs obtained (Table 1) were similar for 400-m

257 Australian and US Olympic swimmers (0.6%); instead, for 800- and 1500-m races the
258 CVs were higher (1.0%) (Pyne et al., 2004) than the results obtained at the European
259 Championships.

260

261 In 400-m, 800-m, and 1500-m swimming events, where there are only two rounds, the
262 swimmers must achieve a balance between an optimal performance to qualify and, at the
263 same time, a reduction of the energy expenditure during these middle and long distance
264 events (Lipińska et al., 2016b; Mauger et al., 2012). Therefore, the pacing strategy
265 adopted may allow to reduce accumulated fatigue and compete in the final with greater
266 guarantees of success. In this sense, successful swimmers present higher CV between
267 rounds to save their best performance for the final (Thompson et al., 2004). Previous
268 studies had established that swimmers must improve by ~1% their performance
269 variability between heats and final (Pyne et al., 2004), as it observed in the results of the
270 present study with medallist and non-medallist finalist (Table 3). The medallist obtained
271 higher CV average between rounds (1.00% for males and 1.08% for females) compared
272 to non-medallist (0.22% for males and 0.47% for females). Hence, it could suggest that
273 best swimmers perform conservative pacing strategies in heats to improve their race times
274 in final.

275

276 The CVs obtained could be affected by the competitive context of the event. In this study,
277 the results of the European Championship were obtained just nine weeks before Tokyo
278 2020 Olympics Games. For this reason, having two major events in a relatively short
279 period, requires coaches and swimmers to program with block periodization where
280 athletes vary their performance peak (Issurin, 2008). Although in some research, a cycle

281 based on 8 to 16 weeks has been considered a typical period in swimming to produce the
282 desired adaptations (Hellard et al., 2017), which could allow the participants of the
283 European Championship to simulate the optimal performance for a major event. Also it
284 has been analysed swimmers' performance over 3-weeks period (I Mujika, Padilla, &
285 Pyne, 2002), 12-month period (Pyne et al., 2004), and even performance progression in
286 major events over 7 years (Mujika et al., 2019). Therefore, the variability analysis should
287 consider the different events that may affects to the performance competition during a
288 specific period.

289

290 The relevance of pacing could be valuable for performance enhancement in a swimming
291 event, since adopting the right pacing strategy might lead to a better competition result.
292 Moreover, the analysis of variability and progressions provide useful information about
293 the smallest changes in performance. It is important to highlight that the results obtained
294 here may be useful both for coaches and swimmers due to the recent inclusion of the
295 men's 800-m and women's 1500-m freestyle in the 2020 Tokyo Olympics Games. For
296 these reasons, the study of pacing strategies, variability, and progression, especially in
297 these two events, provide information for the development of current competitive
298 swimming.

299

300 **CONCLUSION**

301

302 Elite swimmers adopted a parabolic profile and increased their variability and
303 performance progression in 400-, 800-, and 1500-m races during the 2021 European
304 Championship between heats and finals. Medallist swimmers showed higher CV values

305 and progression compared to non-medallist finalist between rounds. In this study, only
306 race times achieved in a major event were analysed; therefore, future research should
307 consider the competitive context, since being an Olympic year, coaches and swimmers
308 could consider the European Championship as a preparatory competition in their
309 schedule.

310

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312

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318

319 **DISCLOSURE STATEMENT:**

320

321 The authors have no conflicts of interest to report.

322

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396

397

398 **TABLES AND FIGURES CAPTIONS**

399

400 **Figure 1.** Pacing profile in heats and finals, coefficient of variation (CV), and relative
401 change ($\% \Delta$) between rounds in the finalist of 400-m races.

402

403 **Figure 2.** Pacing profile in heats and finals, coefficient of variation (CV), and relative
404 change ($\% \Delta$) between rounds in the finalist of 800-m races.

405

406 **Figure 3.** Pacing profile in heats and finals, coefficient of variation (CV), and relative
407 change ($\% \Delta$) between rounds in the finalist of 1500-m races.

408

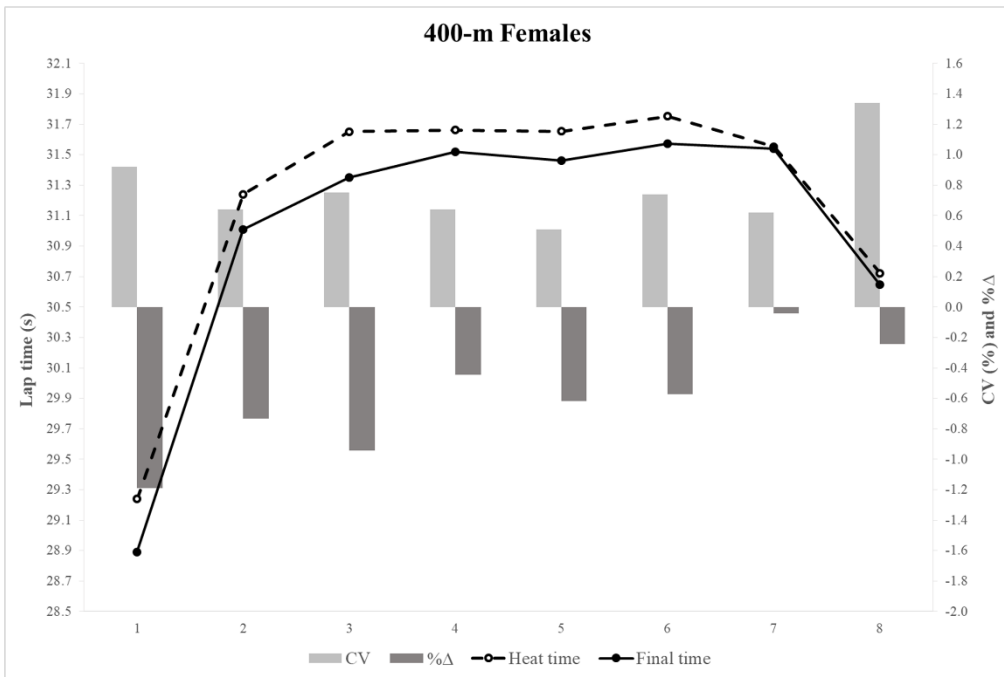
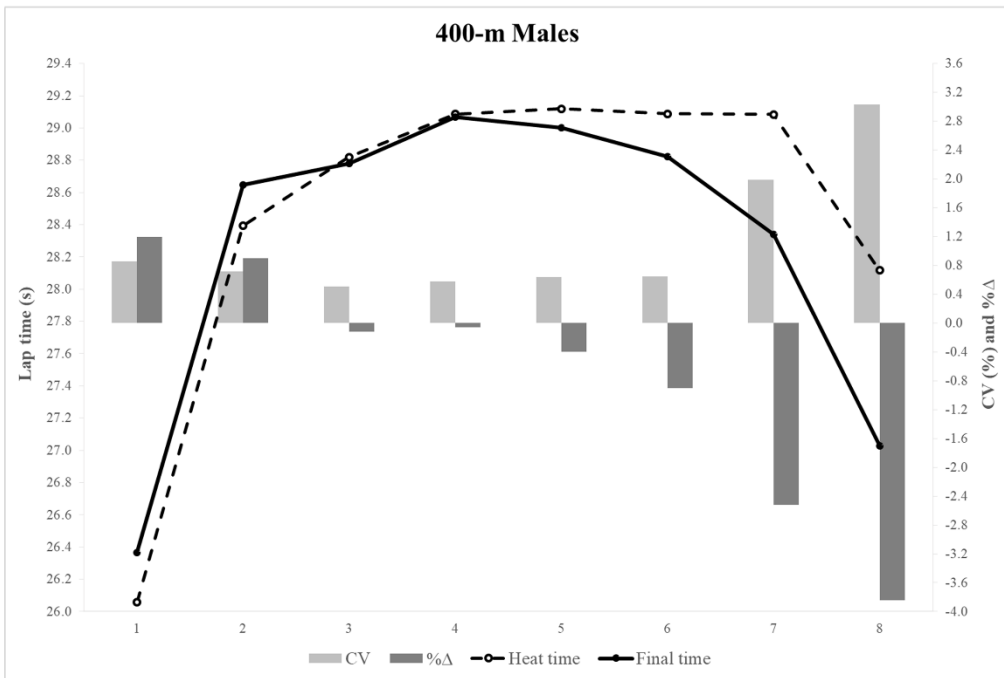
409 **Table 1.** Within-athlete coefficient of variation (CV) and relative change ($\% \Delta$) between
410 heats and final.

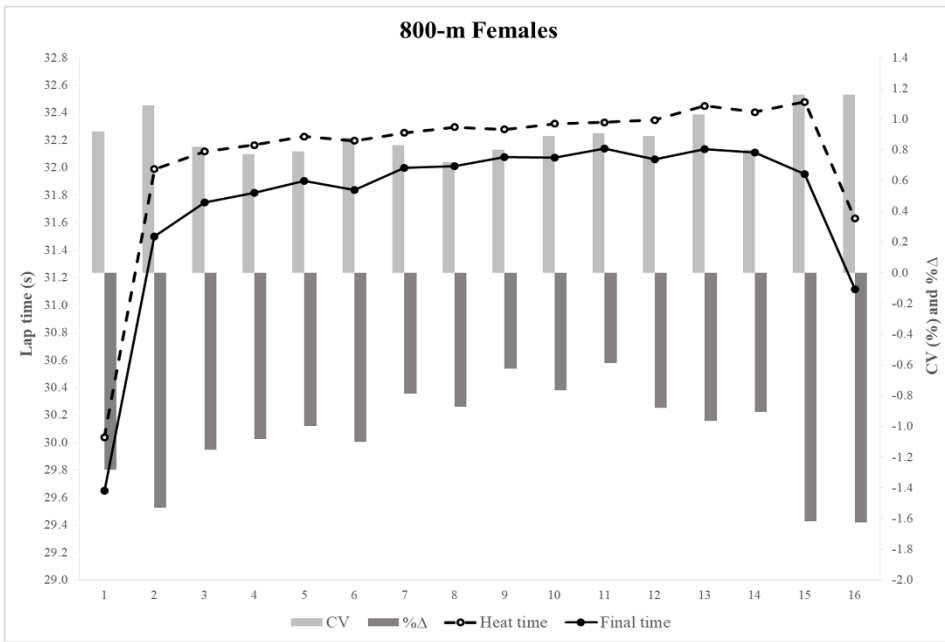
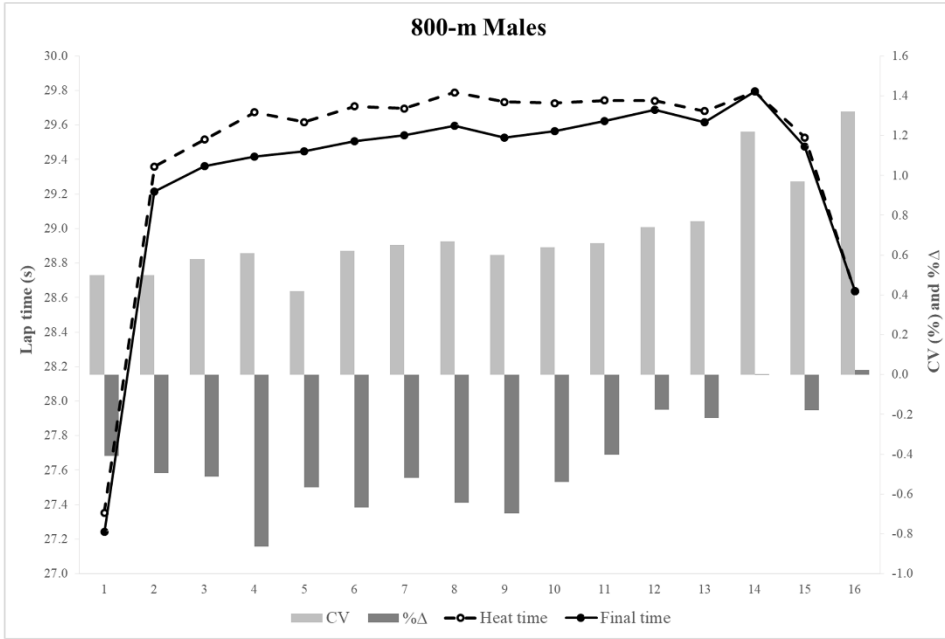
411

412 **Table 2.** Inter-athlete coefficient of variation (CV) in heats and final.

413

414 **Table 3.** Comparison of coefficient of variation (CV) between medallist and non-
415 medallist swimmers.





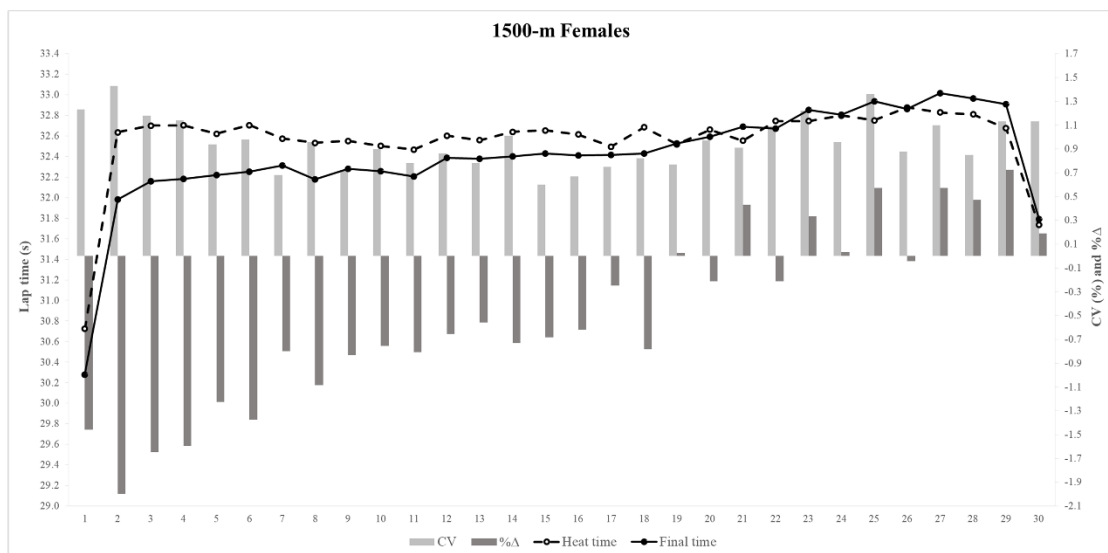
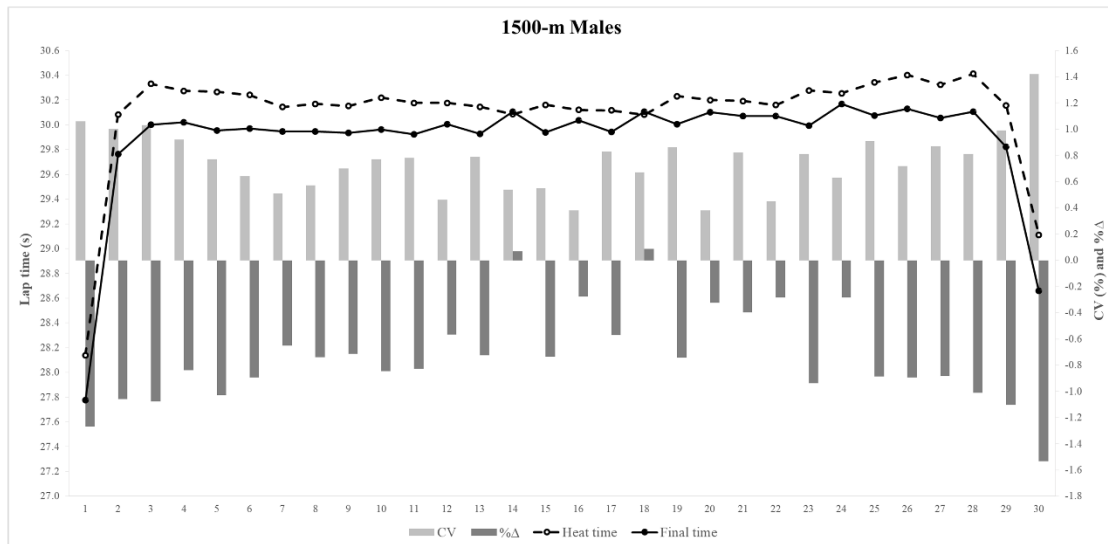


Table 1. Within-athlete coefficient of variation (CV) and relative change (% Δ) between heats and final.

EVENT	Males			Females		
	CV	p	% Δ	CV	p	% Δ
400-m	0.53 \pm 0.40	0.003	-0.75 \pm 0.57	0.65 \pm 0.29	0.038	-0.59 \pm 0.82
800-m	0.47 \pm 0.46	0.073	-0.43 \pm 0.81	0.75 \pm 0.57	0.003	-1.04 \pm 0.80
1500-m	0.55 \pm 0.68	0.030	-0.73 \pm 1.01	0.71 \pm 0.48	0.062	-0.50 \pm 1.15
MEAN	0.52 \pm 0.49		-0.64 \pm 0.80	0.70 \pm 0.45		-0.71 \pm 0.92

Table 2. Inter-athlete coefficient of variation (CV) in heats and final.

EVENT	Males		Females	
	Heats	Final	Heats	Final
400-m	2.9%	0.6%	3.7%	0.9%
800-m	2.6%	1.0%	3.4%	1.0%
1500-m	2.6%	1.2%	2.0%	1.2%
MEAN	2.7%	1.1%	3.0%	1.0%

Table 3. Comparison of coefficient of variation (CV) between medallist and non-medallist swimmers.

EVENT	Males		Females	
	Medallist	Non-medallist	Medallist	Non-medallist
400-m	0.96 \pm 0.20	0.28 \pm 0.21	0.90 \pm 0.36	0.50 \pm 0.12
800-m	0.93 \pm 0.25	0.20 \pm 0.30	1.26 \pm 0.35	0.44 \pm 0.45
1500-m	1.13 \pm 0.80	0.20 \pm 0.29	1.10 \pm 0.20	0.48 \pm 0.46
MEAN	1.00 \pm 0.41	0.22 \pm 0.26	1.08 \pm 0.30	0.47 \pm 0.34