

**The spatiotemporal control of expert tennis players when returning first serves: a
perception-action perspective**

José A. Navia

^aFacultad de Ciencias de la Actividad Física y del Deporte (INEF), Universidad
Politécnica de Madrid, Madrid. ORCID: 0000-0003-1218-5033

Twitter: @naviagallego

Carlos Avilés

^bFacultad de Educación, Universidad Complutense de Madrid, Madrid.

ORCID: 0000-0001-9540-1808

Twitter: @caavilesv

Matt Dicks

^cSchool of Sport, Health and Exercise Science, University of Portsmouth, United
Kingdom. ORCID: 0000-0001-6584-2733

Twitter: @MattDicks7

Luis M. Ruiz

^aFacultad de Ciencias de la Actividad Física y del Deporte (INEF), Universidad
Politécnica de Madrid, Madrid. ORCID: 0000-0002-9678-5986

Twitter: @lmruizperez

Corresponding author: Carlos Avilés.

Rector Royo Villanova, 1, Ciudad Universitaria, 28040 Madrid or caviles@pdi.ucm.es

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Abstract

1 The aim of the current experiment was to examine the spatiotemporal control of expert
2 tennis players while executing first service returns within a representative experimental
3 setting. We recruited and tested twelve male expert tennis players in hard courts. A
4 comprehensive analysis of the timing (eleven temporal variables analyzed at 300 Hz)
5 and performance success of the return actions were carried out, while simultaneously
6 considering task constraints such as the accuracy and the speed of the serves. Temporal
7 organization of return actions were scaled relative to the server's racket-ball contact (5
8 ms), an adaptation of fly-time of the split-step, which resulted in consistent landings
9 (133 ms), and initiation of lateral movements towards the ball –with no response errors–
10 after the server's stroke (around 177 ms). Poorer returns occurred when responding to
11 accurate serves accompanied by late trunk movements towards the ball. Returners
12 scaled the timing of the response to the unfolding action of the serve in order to support
13 both spatial and temporal accuracy. These novel findings highlight the significance of
14 the study of fast-ball sports in representative settings and offer further detail on the
15 spatiotemporal control of skilful perception-action.

16 **Keywords:** Ecological psychology; expertise; sport performance; tennis returns

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23 **1. Introduction**

24 Elite sports are replete with situations where athletes execute highly-skilled
25 movements against competing opponents. The aim of many scientists has been to reveal
26 the possible attributes that underpin such sporting excellence (Williams & Jackson,
27 2019). In particular, researchers have aimed to measure the underlying mechanisms of
28 expertise in fast-ball sports where athletes are faced with situations that offer minimal
29 time for perception-action. As such, examples of commonly studied fast-ball sports
30 include batting in cricket (Mann et al., 2010), baseball (Gray, 2002) and the penalty
31 kick in football (Dicks, Button, et al., 2010b; Navia et al., 2017). One example of a fast-
32 ball situation that has received a large amount of research interest is the return of serve
33 in tennis (Avilés et al., 2019). The specific temporal demands for the player returning
34 ‘flat’ fast serves leaves approximately 700 ms from the point of server racket-ball
35 contact until the ball reaches the other side of the court (Vaverka et al., 2003), yet expert
36 tennis players are still able to return the ball accurately to a remarkably high degree
37 (Triolet et al., 2013). A question of interest for researchers has been to identify whether
38 skilled players are capable of using advance information from an opponent’s
39 movements (e.g., Farrow & Abernethy, 2003), and the game context (Farrow & Reid,
40 2012) in order to accurately predict ball-flight spatial trajectory prior to its onset. A
41 prominent consideration has been whether such early predictions underpin elite
42 performance by providing players with more time to coordinate action (Müller &
43 Abernethy, 2012; Vernon et al., 2018). Indeed, this primary research question has
44 motivated some seminal works in racket sports, which have shown that experts are able
45 to better predict event outcomes more accurately than less-skilled performers
46 (Abernethy et al., 2001).

47 Recent findings have led some researchers to question the assumption that elite
48 athletes do predict spatial ball-flight trajectories well in advance of their onset. For
49 example, through the observational coding of professional tennis matches, Triolet and
50 colleagues (2013) revealed that the frequency of lateral motion of the player's body or
51 racket before the opponent's stroke (see also Mecheri et al., 2019) –even when returning
52 second serves– was marginal. This important observation was further emphasised in a
53 recent systematic review of anticipation of tennis serves (Avilés et al., 2019). These
54 authors reported that studies of expert tennis players consistently demonstrated that
55 lateral movement towards the (expected trajectory of the) ball prior to racket-ball
56 contact rarely occurs under representative conditions when responding to first serves.
57 Hence, the returners' actions towards the ball seem to occur between 160 - 170 ms after
58 the server's stroke (Mecheri et al., 2019; Triolet et al, 2013). Importantly, the results
59 from Triolet and colleagues (2013) and other studies (Mecheri et al., 2019; Shim et al.,
60 2005) have also revealed infrequent response errors (i.e., movement to the wrong spatial
61 location) when elite players initiate their responses after racket-ball contact of the serve.
62 Analysis of the kinematic actions of tennis serves have suggested that specifying
63 information concerning ball trajectory unfolds close to the point of racket-ball contact
64 (Huys et al., 2008). Hence, starting the lateral movement of the returning action before
65 server's ball-racket impact increases the likelihood of a decrease in spatial accuracy.

66 Such findings have helped to inform the contemporary proposal that skilled
67 performers use kinematic information not only to spatially adapt their movements to the
68 trajectory of the ball but also to adjust the timing of the return action (Dicks et al., 2019;
69 van der Kamp et al., 2018). Thus, it is proposed that expertise may not manifest itself as
70 early predictive judgements that offer a maximum time period to execute an action, but
71 in contrast, skill can be conceptualised as the accurate calibration of actions that are

72 scaled to the spatiotemporal demands of the task (Fajen, 2005). Hence, the accurate
73 scaling of one's own action capabilities (e.g., agility) to the spatiotemporal demands
74 (e.g., time available) provides a maximum time period before initiating the response in
75 order to use more reliable information. By way of example, Mecheri and colleagues
76 (2019) reported that tennis players who were more powerful – and therefore could cover
77 an equivalent distance in a shorter time – waited later before initiating their actions and
78 produced more consistent returns of serve (see also Dicks, Davids et al., 2010). Such
79 results add further empirical support to the suggestion that elite performance is not only
80 predicated on spatial anticipatory behaviour (Avilés et al., 2019) but that experts
81 prospectively control their actions, utilising unfolding information via affordance based
82 control (Fajen, 2005). However, despite the efforts made by researchers to investigate
83 the spatiotemporal control of actions during serve returns (Avilés et al., 2014; Mecheri
84 et al., 2019), the extent to which highly-skilled performance can be explained by the
85 accurate coupling of return actions that are timed relative to service actions is still
86 limited, as the spatiotemporal constraints of the tasks have not been simultaneously
87 considered. To this end, research suggests that return performance is influenced by both
88 the speed and the accuracy of serves (Haake et al., 2000; O'Donoghue & Ballantyne,
89 2004). Faster serves reduce the time to contact and serves in the corner of the service
90 court increase the distance to reach the ball, so both shape the time available to
91 complete the return action (Navia et al., 2017).

92 The purpose of this study was to build on existing literature (e.g., Mecheri et al.,
93 2019) and further current understanding of the spatiotemporal control of expert tennis
94 players in a competitive training situation, that simulated match-play conditions. We
95 sought to do so whilst considering, for the first time in literature, the speed and accuracy
96 of the serve as a covariate for timing and performance. In addition, in order to offer

97 unique experimental rigour, the timing of actions was analysed with a detail of 300 Hz,
98 which allowed us to accurately quantify the spatiotemporal control of actions. Thus, the
99 current study addressed the recommendations provided in Avilés et al (2019) recent
100 systematic review of anticipation in tennis. In line with previous findings (Mecheri et
101 al., 2019; Shim et al., 2005; Triolet et al., 2013), players were expected to initiate the
102 split-step and lateral displacement movements after the racket-ball contact as a result of
103 scaling to the opponent's unfolding action (Navia et al., 2017). We also hypothesized a
104 decrease in player performance when facing faster and more accurate serves (van der
105 Kamp et al., 2018) given the increased spatiotemporal demand.

106 **2. Method**

107 **2.1 Participants**

108 Twelve right-handed male expert tennis players (age: 21.4 ± 5.5 years old)
109 volunteered to participate in the study. Based on the International Tennis Federation
110 (ITF), International Tennis Number (ITN) ranking criterion (ITF, 2004) all players were
111 categorized at ITN 1 or 2, with extensive professional tournament experience ($12.83 \pm$
112 6.3 years). Six of the participants were established Association of Tennis Professionals
113 (ATP) players, with two ranked in the top ten for male singles in the season(s)
114 preceding or following data collection. The remaining six players were highly ranked
115 nationally within Spain. All participants were undertaking extensive regular training at
116 the time of the experiment (21.7 ± 6 h/week). The participants acted as both servers and
117 returners during the experiment. Before testing, all participants provided written consent
118 and ethical approval was obtained from the local University's ethics committee.

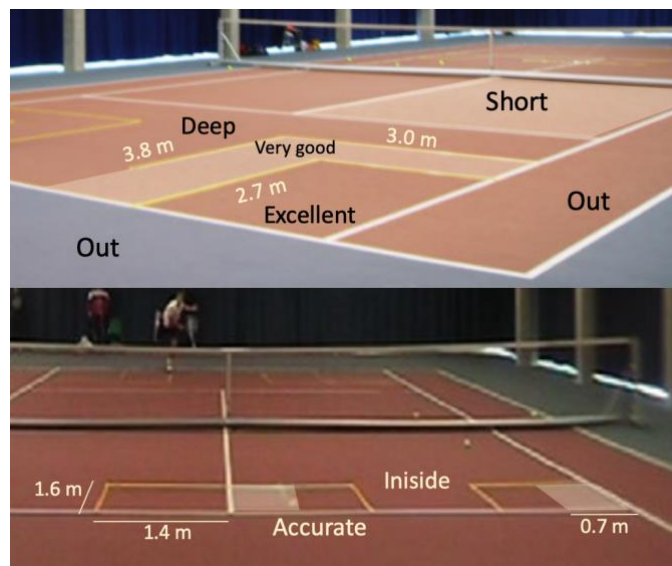
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121 **2.2 Material**

122 Personal racquets of participants and Dunlop Fort All Court tennis balls were
123 used during testing. The timing of the serve and return actions were recorded at 300 Hz
124 using a high-speed camera (Casio Exilim F1), located behind the receiver ~4.5 m from
125 the baseline. Two 25Hz cameras (Cannon MV950 and JVC GY-301E) situated 5 and 4
126 m behind the baseline captured the serve and return locations, respectively. Speed of the
127 serve was registered with a radar (Sports Radar SR3600), placed at a height of 2.9 m
128 and 2.9 m behind the server, and connected to a laptop that displayed and recorded the
129 velocity of the ball (Sports Radar Data Acquisition Software V.1.1. Technology). Tape
130 was used to mark target areas for the serves and the different areas of the return
131 (Hornery et al., 2007). Recordings from the cameras were analyzed using Kinovea
132 V.0.8.7. and then coded into an excel sheet.

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135 **Figure 1.** Areas of the return (top) and targets for the serve (bottom).

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139 **2.3 Procedure and design**

140 All testing was carried out on a tennis hard-court. Prior to testing, all participants
141 undertook a warm-up that consisted of 5 minutes of jogging and stretches, 10 minutes
142 of tennis rallies, and 5 minutes of specific serve and return actions (i.e., familiarization).
143 The experiment consisted of a period of simulated match-play, during which, each
144 participant carried out 50 returns. After each set of 10 trials, players changed the side of
145 the court (3 sets at deuce and 2 at advantage). Testing was organized in pairs of
146 participants in such a way that one player started serving and another player returning.
147 Roles were interchanged (server-returner) after the 3rd, 6th and 8th set of 10 trials until
148 the completion of 100 trials (i.e., 50 returns for each participant). Players were told to
149 follow their habitual routine and perform exactly as they would do during serves and
150 returns. There was a mean period of 13 seconds between trials –a time which is within
151 the stipulated time frame in tournaments–, meaning participants had regular periods of
152 rest during testing. Moreover, participants had 2.5 minutes break between sets. In total,
153 testing took between 45 and 60 min for a pair of players. Tennis players were
154 encouraged to perform flat serves as fast as possible towards the targets placed in T and
155 wide areas (Mecheri et al., 2019). The direction of serve was randomized and equally
156 distributed (T and wide) and servers were provided with instruction of where to aim
157 before each trial. Returners –who did not know the direction of the serves– were
158 encouraged to return as best as possible.

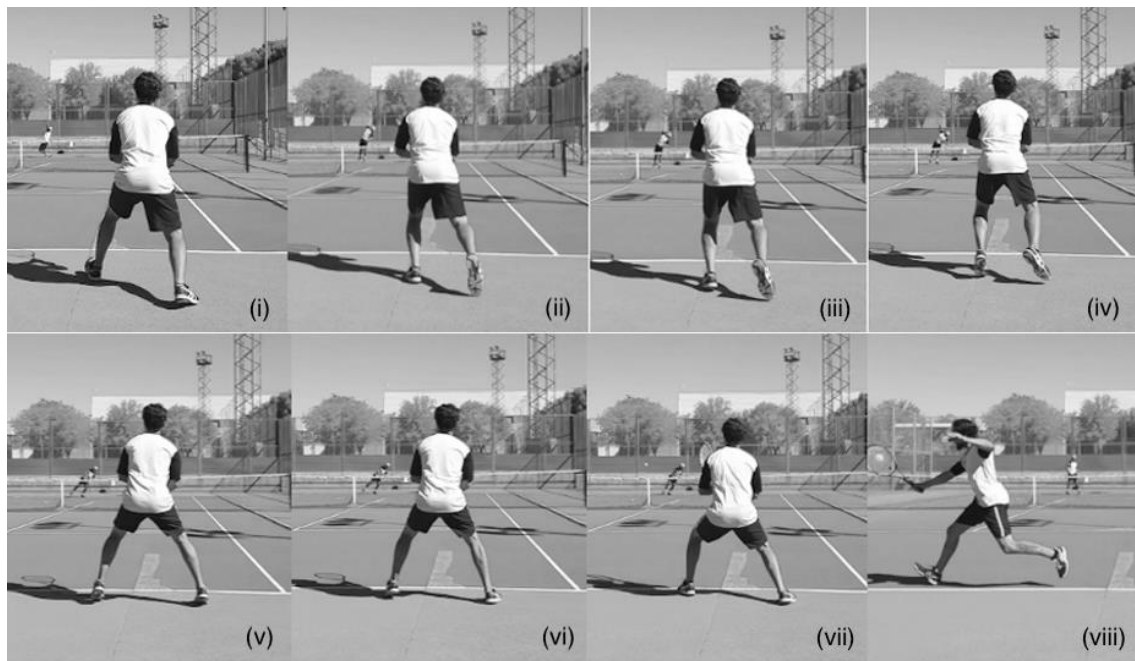
159 **2.4 Measures**

160 A total sample of 600 trials were obtained across all participants. 249 of the
161 serves (41.5%) were directed out of the service box and were therefore not used for
162 analysis. In addition to the side of the court (deuce vs advantage) that the serve was hit,

163 and serve placement (T or wide), the speed of the serve was also registered for each
164 trial. Return performance and serve accuracy were assessed from the video recordings
165 (Figure 1). Following prior research (Avilés et al., 2014; Hornery et al., 2007), a scale
166 was applied to returns according to the landing location of the ball (Figure 1 top): 0
167 points for an ace, 1 for a short, 2 for a deep, 3 for a very good, and 4 for an excellent
168 return. Using a software grid, accurate serves were registered when the ball bounced in
169 the cornered half of the target (see Figure 1, bottom). Some behavioral measures of the
170 returner were also coded: (i) whether or not a player moved before the server's racket-
171 ball impact (i.e., directional split-step, see below); (ii) whether the player moved in the
172 same direction as the serve (i.e., movement direction); and (iii) whether the player's
173 foot landing after the split-step was congruent or incongruent with the ball direction
174 (Avilés et al., 2014).

175 In order to add further detail to previous research (Avilés et al., 2002; Filipcic et
176 al., 2017; Gillet et al., 2010), nine temporal events of the returning players were coded
177 relative to the server's stroke. Negative values denoted events before server's racquet
178 ball-contact and positive values were recorded for events after racquet-ball contact. As
179 depicted in Figure 2, preparatory actions involved: (i) Heel take-off of the back leg; (ii)
180 Impulse (when the lumbar area –representing the center of gravity– moved upwards);
181 (iii) Heel take-off of the front leg; (iv) Split-step (front foot take-off); (v) First foot
182 landing – toe contact – following the split-step; and (vi) Second foot landing – toe
183 contact – following the split-step. (vii) Trunk onset was defined as the first observable
184 movement of the trunk, arms or racquet, towards the actual ball trajectory. Finally, (viii)
185 Return indicated the moment of the return stroke (or ball passing by the returner in the
186 case of an ace). In some of the aces, the returner did not try to contact the ball after
187 being aware that ball was far away from their range, so the return was not coded in such

188 instances ($N = 37$). (ix) Directional split-step was also coded only in the instances ($N =$
189 36) when the player initiated a lateral split-step –correct or not– with any part of the
190 body towards either side of the court before the server’s impact. Two additional
191 spatiotemporal variables were computed: (x) Fly time –from the initiation of the split-
192 step to the first foot landing; and (xi) Movement time –from Trunk onset to Return
193 (Avilés et al., 2002; Avilés et al., 2014; Vaverka et al., 2003). A second researcher
194 analyzed 13 variables from 20 randomly selected trials, obtaining a global Cronbach’
195 alpha = 0.94.
196



197

198 **Figure 2.** Events coded in during the sequence of the return: (i) back heel take off; (ii)
199 impulse; (iii) front heel take-off; (iv) split-step; (v) first foot landing; (vi) second foot
200 landing; (vii) trunk onset; and (viii) return stroke.

201

202 2.5 Statistics

203 Due to the hierarchical nature of collected data (i.e., residuals of observations are
204 not independent), multilevel linear models were applied: Level 1 trials, Level 2
205 participants. Initial exploratory analysis included normality tests (K-S, skewness and

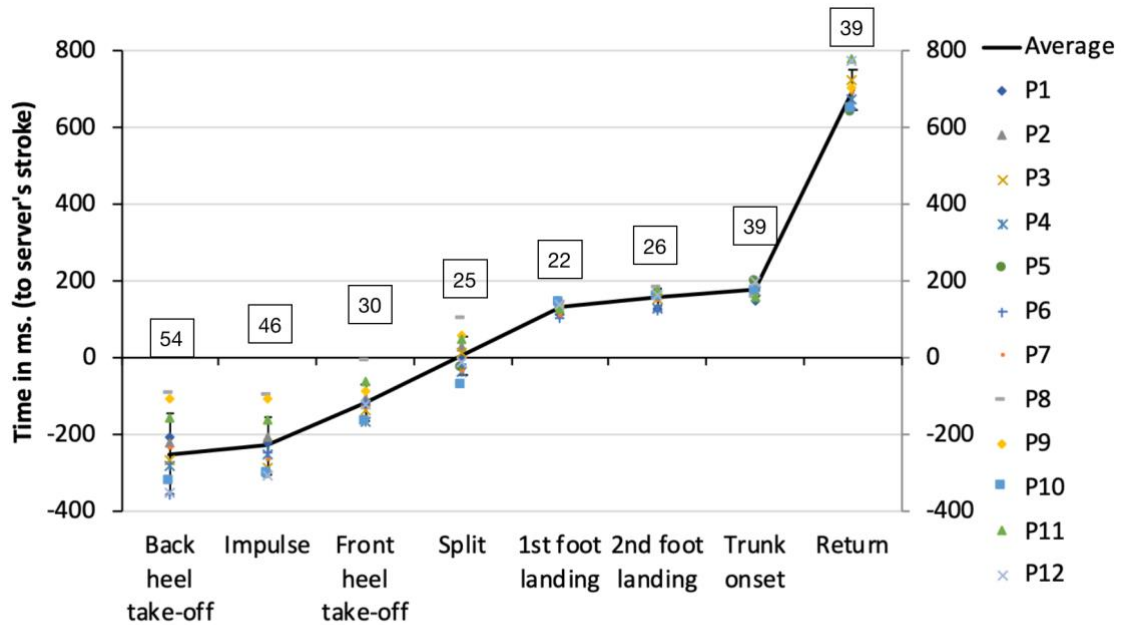
206 kurtosis) and visual inspection (histograms and Q-Q plots). After descriptive and
207 correlational analysis, we first tested in mixed models random intercepts with fixed
208 slopes, then random intercepts and slopes, and finally examined interactions between
209 fixed factors whenever possible (Field, 2018). Row scores were used as predictors
210 (without centering). After the models, Q-Q and scatterplots checked normality and
211 homoscedasticity. Effect size for Bonferroni post-hoc were expressed in d , with values
212 of 0.2, 0.5 and 0.8 for small, medium and large effect, respectively (Cohen, 1988). The
213 level of significance for all analysis was set at $\alpha = 0.05$. Using a within factor repeated
214 measures ANOVA test provided.
215 Using a within factor repeated measures ANOVA test provided by G*Power 3 (Faul et
216 al., 2007) a post-hoc power of 0.83 ($1-\beta$) was computed considering a medium effect
217 size ($\eta_p^2=0.06$), $\alpha = 0.05$, and 50 trials per 12 participants. IBM SPSS V.25 (Armonk,
218 NY: IBM Corp., USA) and R-based software jamovi V.1.6.1 (www.jamovi.org) were
219 employed for statistical analysis and figure production.

220 **3. Results**

221 The mean service speed was 176 km/h ($Mdn = 176$, $SD = 10$). The entire
222 temporalization of the return (from back heel take-off to return stroke) took slightly less
223 than a second (Figure 3). Return actions followed this temporal sequence: heel take-off
224 of the back leg ($M = -253$ ms, $Mdn = 250$, $SD = 116$), impulse ($M = -228$ ms, $Mdn = -$
225 233 , $SD = 83$), heel take-off of the front leg ($M = -117$ ms, $Mdn = -117$, $SD = 56$) and
226 split-step ($M = 5$ ms, $Mdn = 0$, $SD = 53$). The average fly-time was 126 ms ($Mdn = 123$,
227 $SD = 50$) before landing, first contacting with one foot (first foot landing, $M = 131$ ms,
228 $Mdn = 133$, $SD = 27$) and then with the second foot ($M = 156$ ms, $Mdn = 150$, $SD = 33$).
229 On average, tennis players moved towards the ball 177 ms after the server's stroke (i.e.,

230 trunk onset: $Mdn = 176$, $SD = 42$). The mean time of the return strokes was 692 ms
 231 ($Mdn = 678$, $SD = 67$), which represented an average movement time of 519 ms ($Mdn =$
 232 509, $SD = 77$).

233



Timing of the return

234
 235 **Figure 3.** Individual profiles of participants' temporalization of returns. Mean intra-
 236 participant variability (ms) is displayed in squares as the average of individual SD's.
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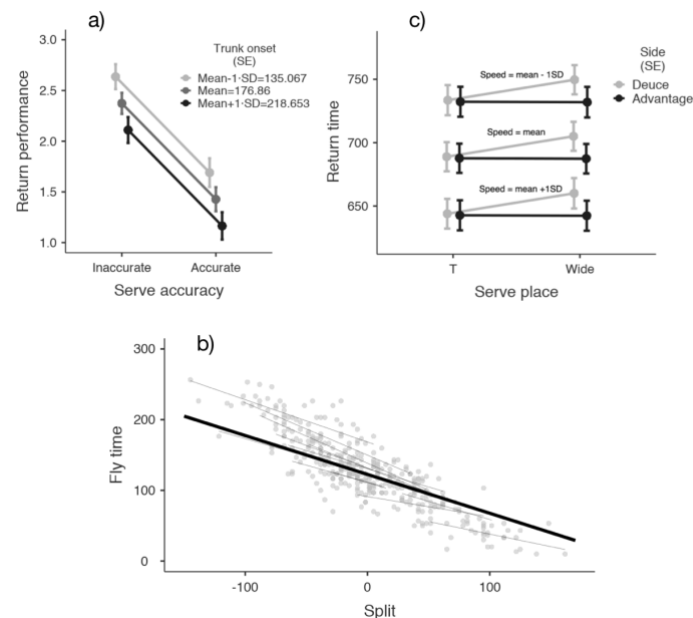
238 In most of the trials (89.8%), players performed a neutral split-step (a forward
 239 movement), followed by a trunk movement towards the ball. In 10.2% of the trials
 240 players initiated a directional split-step (lateral movement towards either side) before
 241 the completion of the server's action ($M = -122$ ms, $Mdn = -115$, $SD = 84$ ms), followed
 242 by a trunk movement towards the actual direction of the ball. Mixed-model analysis
 243 (random intercept and fixed slope, $BIC = 3614.07$, $ICC = 0.11$) did not reveal any
 244 differences in Trunk onset (i.e., actual movement towards the ball) between trials
 245 preceded by directional split-step (estimated marginal mean of Trunk onset = 182 ms)
 246 and neutral split-step (estimated marginal mean of Trunk onset = 176 ms), *estimate =*

247 5.47, $F(1,350.79) = 0.57$, $p = 0.452$. However, returners moved towards the wrong
248 direction significantly more often when they carried out a directional split-step (12 out
249 of 36 cases) than when a neutral split-step was performed (2 out of 313); as confirmed
250 by mixed-model analysis (random intercept and fixed slope, $BIC = 87.64$) $estimate = -$
251 4.36, $z = -5.5$, $\chi^2(1) = 30.26$, $p < 0.001$.

252 From the 351 trials, 207 (59%) were performed from the deuce side of the court
253 and 144 (41%) were performed from the advantage side. Considering the direction of
254 serves, 179 (51%) were directed towards the T (central) and 172 (49%) were directed
255 wide (Supplemental 1). Performance of the returns were similar for both sides of the
256 court and serve directions, as all p 's > 0.05 in main effects and interactions. Return
257 success was primarily influenced by the accuracy of the serve (145 trials landed in the
258 'accurate' area and a further 206 landed in the 'inside' area, see Figure 1 bottom).
259 Unlike serve speed, the model significantly improved when the time of trunk onset was
260 added (Figure 4a), indicating that an earlier trunk onset facilitated better responses.
261 Hence, the best final model (random intercepts, fixed slopes, $BIC = 1199.89$, $ICC =$
262 0.02) that explained the results included serve accuracy, $F(1, 351) = 45.89$, $estimate = -$
263 0.95, $p < 0.001$ and trunk onset, $F(1, 309) = 14.07$, $estimate = -0.006$, $p < 0.001$.

264 Preparatory actions and the initiation of the return action were not temporally
265 constrained by the speed of the serve, as no main effects for speed were found on
266 impulse, split, fly-time, foot landings or trunk onset (all p 's > 0.05). Rather, results
267 revealed that players regulated their split-step relative to information from the server's
268 impact of the ball, as indicated by the results of three variables. First, the split-step (i.e.,
269 last foot tip take-off) was tightly coupled with racquet ball-contact ($M = 5$ ms, $Mdn = 0$,
270 $SD = 53$). Second, as Figure 4b depicts, players adjusted their fly time, so that later
271 splits were followed by earlier first foot landings, mixed-model (random slopes and

272 intercept, $BIC = 3138.15$, $ICC = 0.65$), $F(1,11.2) = 74.81$, *estimate for intercept*=
 273 122.25 , *estimate for split* = -0.55 , $p < 0.001$. Third, first foot landing occurred
 274 significantly more often with the opposite foot (60%) compared to the same foot (40%)
 275 as the direction of the ball, (random intercept, $BIC = 480.29$), $z = 2.55$, *estimate* = 0.41 ,
 276 $p = 0.011$. Actually, mixed model (random slopes and intercept, $BIC = 3588.41$, $ICC =$
 277 0.13) revealed an earlier trunk onset when landing with the opposite foot first (estimated
 278 marginal mean of Trunk onset = 170 ms) than when landing with the same foot first
 279 (estimated marginal mean of Trunk onset = 190 ms), *estimate* = -19.93 , $F(1,10.52) =$
 280 7.23 , $p = 0.022$.



281
 282
 283 **Figure 4.** Performance of tennis players. a) Return performance as a function of accuracy
 284 of the serve and lateral trunk and/or racquet lateral initiation times; b) Relationship
 285 between split moment and fly time during the split-step. Solid line and fine lines represent
 286 regression model and individual random effects, respectively; c) Times of the return
 287 stroke by side, place and speed of the serve.

288 While speed had no effect on return performance, impulse, split, fly-time, foot
 289 landings or trunk onset, it was observed that later return strokes occurred when
 290 responding to slower serves, and also when returning wide serves that were executed
 291 from the deuce side of the court (Figure 4c). Mixed-model (fixed slopes and random

292 intercepts, $BIC = 3044.41$, $ICC = 0.68$) revealed a main effect for speed, $F(1,314) =$
293 261.31 , $estimate = -4.44$, $p < 0.001$, for side, $F(1,299,02) = 7.40$, $estimate = 17.69$, $p =$
294 0.007 , serve place, $F(1,298.53) = 5.40$, $estimate = 0.33$, $p = 0.021$, and an interaction
295 between speed and serve direction, $F(1, 299.64) = 5.30$, $estimate = -16.45$, $p = 0.022$.
296 Post-hoc tests with Bonferroni correction revealed that return time of Deuce-Wide
297 serves significantly differed from Deuce-T, $t(308) = -3.44$, $d = 0.39$, $p = 0.003$, and from
298 both Advantage T and Wide, $t(307) = -3.73$, $d = 0.42$, $p < 0.001$ and $t(308) = 3.47$, $d =$
299 0.39 $p < 0.001$, respectively. No further pairwise comparisons were significant.

300 **4. Discussion**

301 The aim of the current study was to examine the spatiotemporal control of expert
302 tennis players when facing first serves (Avilés et al., 2019). In line with recent studies
303 (Mecheri et al., 2019; Triolet et al., 2013) players were found to accurately calibrate their
304 split-step and lateral displacement movements to the opponent's unfolding actions.
305 Preparatory movements (i.e., heel take-off of the back leg, impulse and heel take-off of
306 the front leg) occurred well before serve release –from -253 ms onwards– whilst,
307 measures revealed a relatively high variation between and within participants. In contrast,
308 during the time-course of the serving action, there was a notable coupling between the
309 split-step and the server's stroke, as players consistently performed the split-step at the
310 time of racquet-ball contact (around 5 ms). As previously reported (Mecheri et al., 2019),
311 the timing of the landing action was very consistent. During the split-step, a systematic
312 regulation of the fly time length was observed as players landed at approximately the
313 same time just after ball release (131 and 156 ms for first and second foot landing,
314 respectively). Furthermore, landings after the split-step, occurred slightly but
315 significantly more frequently with the contralateral foot to the ball direction. This
316 behaviour reflects a coordinative adaptation to support the following lateral displacement

317 towards the ball (Avilés et al., 2014; Elliot et al., 2009), which occurred 177 ms after
318 racquet-ball contact. A future research challenge is to examine the actual prevalence of
319 contralateral foot landings –and subsequent behaviours– with respect to expertise and/or
320 individual capabilities (e.g., agility) and in changing contexts (e.g., court type).

321 In comparison with extant literature (Shim et al., 2005; Triolet et al., 2013), one
322 particular advance of the current research is the examination of service speed relative to
323 the spatiotemporal control of serve returns. One might expect that the speed of the serve
324 would not have an impact upon the timing of return actions before (heel take-offs,
325 impulse) or around (split-step) the service racket-ball contact, as information pertaining
326 to ball speed is not available –and predicting the speed of the serve based on server’s
327 preparatory actions seems unlikely given their similarity among trials. A question that
328 remained unsolved was whether the speed of the serve somehow affected the timing of
329 the return actions that take place after the serve. Results indicated that fly time, foot
330 landings, and moving of trunk onset were not significantly influenced by the ball speed.
331 This finding suggests that elite tennis players utilise information from the movement
332 kinematics of the server, in order to regulate the temporal control of the return action
333 (Mecheri et al., 2019). Speed of the serve only influenced the time of the return stroke,
334 meaning that slower serves resulted in players taking more time to cover the court before
335 completing the return action (i.e., the ball took longer to arrive at the racket). Contrary to
336 suggestions in the literature (O’Donoghue & Ballantyne, 2004), speed of the serve in this
337 experiment was not found to determine the returners’ success. Serves in the current
338 experiment were consistently executed at considerable speed ($M = 176$ km/h) given that
339 participants were performing at an expert level and that all serves were executed as first
340 serves. Subsequently, the key determinant for the return success appeared to be the
341 accuracy of the serve, which could be due to the trajectory followed by accurate serves

342 (directed to the corners of the service court). A later initiation of lateral action towards
343 the ball (trunk onset) also resulted in poorer performance, as it reduced the temporal
344 margin to perform the returning stroke (van der Kamp et al., 2018). This result suggests
345 that returners might perform outside a safe spatiotemporal boundary (Fajen & Devaney,
346 2006). That is, when attempting to return serves hit to the corners of the service court,
347 one might have expected that the tennis players would have moved earlier to enable
348 adequate scaling of their response relative to their action capabilities (Dicks, Davids et
349 al., 2010, Navia et al., 2017). However, such earlier scaling would have led to less
350 accurate responses (Shim et al., 2005) as an increased number of movement corrections
351 occur with an earlier directional split-step.

352 Collectively, these novel findings lend empirical support to the proposal that the
353 temporal control (when to move) of real-time interceptive actions are regulated in tandem
354 with the spatial control (where to move) of actions (Dicks et al., 2019). This interpretation
355 is supported by the findings of Navia and colleagues (2017), who revealed that world-
356 class futsal goalkeepers adopted a gaze pattern when saving penalty kicks that was
357 commensurate with the pick-up of information for the temporal control of the saving
358 action. Specifically, this study reported that information exploited prior to penalty taker
359 foot-ball contact appeared to inform on the time to contact between the penalty taker and
360 the ball, rather than spatial anticipation. Information that subsequently unfolded around
361 and following contact (kicking action and ball flight) appeared to subsequently guide the
362 spatial control of the interceptive action, a proposal which is supported by the results from
363 the current study. Hence, tennis players calibrated the return in order to support the spatio-
364 temporal control of movement rather than extending the temporal margin (which led to a
365 reduced response accuracy). Thus, the temporal calibration findings of goalkeepers facing
366 penalties in soccer (Dicks, Davids, et al., 2010) or futsal (Navia et al., 2017), and the

367 apparent inaccuracy of scaling of actions might be actually indicative of an adaptive
368 behaviour.

369 As considered, one concern among tennis researchers and professionals is the
370 extent to which expert players anticipate the direction of the ball under representative
371 conditions (Avilés et al., 2019; Triolet et al., 2013). Results indicated that in 90% of trials,
372 players did not exhibit any lateral movement before the completion of the serve (i.e.,
373 neutral split-step towards the front). Importantly, the examination of actual lateral
374 movement of the trunk and/or racquet towards the ball (i.e., trunk onset) revealed that
375 players did not obtain any performance advantage when a lateral split-step was performed
376 prior to the racket-ball contact in the service action. Conversely, when lateral movement
377 initiation did occur prior to racket-ball contact, this led to an increase in the number of
378 response corrections (Dicks, Button, et al., 2010a). The observation that moving too early
379 led to an increase in the number of response corrections is not a trivial issue in tennis.
380 There is growing evidence that has identified the importance of perception-action as a
381 contributory factor toward reducing the risk of anterior cruciate ligament (ACL) injuries
382 (Collins et al., 2016). To this end, the current results suggest that a facet of skill that may
383 reduce ACL injury risk is an accurate scaling of actions relative to critical events in fast-
384 ball sports. Thus, a requirement for future research is to better understand the role that
385 perception-action can play in sport as a means of reducing injury risk. To this end, a
386 particularly fruitful avenue may be to develop training interventions that focus on
387 improving the timing and calibration of actions (Dicks et al., 2015).

388 **5. Conclusions**

389 The current experiment provides a detailed examination of the performance (service-
390 return action, 11 variables analysed at 300 Hz., spatiotemporal constraints considered) of

391 expert tennis players. In line with previous research (Mecheri et al., 2019), the results
392 indicate that players regulate the timing of returns relative to the time course of the
393 opposing serve action. Return success was determined by the trade-off between
394 regulating actions to ensure response accuracy (spatial control), while optimizing the
395 margin of time to act (temporal control) (Navia et al., 2018). Future research should
396 explore the control of actions as a function of both individual action capabilities (Dicks,
397 Davids, et al., 2010) and changing spatiotemporal demands of the task (e.g., court type,
398 quality of server, etc.). A further step in the understanding of skilful performance should
399 also involve the investigation of the interaction between different sources of information
400 (e.g., preferences of the server) in the control of actions (Cañal-Bruland & Mann, 2015;
401 Navia et al., 2013; Vernon et al., 2018). For instance, whether awareness of information
402 pertaining to an opponent's serving pattern impacts upon the spatiotemporal control of
403 the return.

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411 **Disclosure statement**

412 The authors report no conflict of interest.

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