

This is an Accepted Manuscript of an article published by Taylor & Francis in International Journal of Performance Analysis in Sport on 29/11/23, available at:

<https://www.tandfonline.com/eprint/P35HSMNQMD6PGTN3X9Z/full?target=10.1080/24748668.2023.2288491>

1 **PICKING THE LENGTH: INVESTIGATING HOW BOWLING**
2 **LENGTH INFLUENCES BATTER DECISION-MAKING IN**
3 **INTERNATIONAL MEN’S 50-OVER CRICKET**

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28

29 **Abstract**

30 In this study, 19,587 balls bowled by fast bowlers across two One-Day International
31 (ODI) cricket tournaments were analysed, with the aim of exploring the relationship between
32 bowling lengths and foot-based batter decision-making. Initially, a Chi-Square test of
33 independence was used to determine if bowling lengths were associated with the foot-based
34 decision-making of batters. Subsequently, the relationship of specific foot-based strokes with
35 different bowling lengths were further examined through a Logistic Regression. Post-Hoc
36 analysis of standardised residuals from Chi-Square tests suggest an association between foot-
37 based decision making and dot balls ($\chi^2(1) = 99.798, p < .001$). Furthermore, Logistic
38 Regression results reveal that all length zones except the half-tracker length show statistically
39 significant association with front-foot and back-foot strokes. Respective coefficients of these
40 length zones also show an organisation of specific strokes with front-foot stroke dominating
41 lengths close to the batter (yorker, full-toss, half-volley and good length) and back-foot strokes
42 dominating zones further away from the batter (back of a length and short length). The results
43 confirm interdependence between batter-bowler performances and suggest that batters in ODI
44 cricket prefer meeting the ball on the front foot in their search of scoring runs.

45

46 Keywords: Decision-Making, Bowling Length, Cricket, Cricket Batting, Flexible Decision-
47 Making.

48

49 **Introduction**

50 The dynamic interceptive action of optimally intercepting a ball in cricket batting
51 requires high levels of decision-making performed in considerably little amount of time and it
52 necessitates accurate anticipatory skill as well as the pick-up of advanced information (Müller
53 et al., 2006; Weissensteiner et al., 2009). While each format of cricket involves its own game-
54 specific stipulations, in limited-overs cricket the aim is to score more runs than the opponent
55 within a stipulated number of balls (Mehta & van der Kamp, 2021). With 11 players
56 formulating a team, bowlers are tasked with taking all 10 wickets (when a batter gets out or is
57 dismissed, leading to be replaced by the next batter in the team) to end an innings or finishing
58 the required balls to be bowled, whilst conceding as few runs as possible. A team tries to take
59 wickets by using a combination of bowlers who use different bowling styles, bowling hands,
60 bowling speeds as well as subtle variations in trajectories to get batters out. (Jamil et al., 2022,
61 2023; Mehta et al., 2022). At the highest level of the sport, batters can face a ball regularly
62 traveling over 140 kilometres per hour, which affords a temporal tolerance of only between 2
63 and 5 milliseconds to optimally intercept the moving ball (Sarpeshkar & Mann, 2011; Regan,
64 1997). Additionally, skilled bowlers at the highest level will also enforce trajectory deviations
65 after ball-release such as swinging the ball in the air (conventional or reverse swing) as well as
66 seaming the ball after it bounces or pitches (moving towards or away from the batter), thereby
67 increasing the difficulty of the interceptive task even further (Mehta et al., 2022). These
68 trajectory modifications either pre- or post-bounce lead to late deviations in the ball trajectory
69 and require stroke adjustments in extremely short amounts of time (R. A. Stretch et al., 2000).
70 Such task constraints alongside ever-changing environmental constraints like pitch and ball
71 deterioration requires batters to exhibit functional variability in their motor responses and
72 accuracy in their decision-making (Connor et al., 2020; Phillips et al., 2010, 2012),
73 consequently making the sport an ideal vehicle to study this variability. Batters must use

74 numerous informational cues to determine their optimal response or batting stroke that range
75 from anticipating point of ball release, predicting ball-bounce location or bowling length as
76 well as picking up other specifying cues from the environment. These could include specific
77 positioning of the fielders to gauge the upcoming ball (Müller et al., 2006), postural cues in
78 bowler delivery style (Williams & Jackson, 2019), as well as specific match situations (such
79 as powerplay and death overs) (Jamil et al., 2023). While these informational cues have been
80 studied in isolation, they have not been extensively examined from the perspective of
81 interactions between batter and bowler, especially not with elite competitors performing in
82 their natural environment.

83 Regardless of the number of informational cues, or the quality of these, they must all
84 be used by batters to produce a singular batting stroke. While there are over 25 different batting
85 strokes (based on human annotations of stroke types commonly used in performance analysis),
86 most of them are classified as front-foot or back-foot strokes based on the batter's gross body
87 movement - whether the batter moves forward or backward from their original stance (Khan et
88 al., 2017). These front-foot or back-foot strokes have been studied in relation to the
89 biomechanical and kinematic determinants of individual strokes such as the front foot drive
90 (Stretch et al., 1998; Peploe et al., 2014), but have not been studied together in an interactive,
91 competitive environment where they serve as decision-driven actions. Moreover, there is also
92 a clear dearth of research within decision-making literature relating to the study of discrete
93 decisions taken at the highest competitive level of team sports (Araújo et al., 2019). This paper
94 attempts to examine whether batter decisions elicited in front-foot or back-foot strokes is
95 related to, or influenced by, the bowling length (ball-bounce location) of the deliveries they
96 face. By using the discrete nature of balls bowled in cricket matches, and analysing front and
97 back-foot strokes played by top international cricket batters, this paper intends to fill the gap
98 in literature regarding interactive decision-making at the elite level of performance.

99 Accurate pick-up of information regarding bowling length or ball bounce location is
100 considered a fundamental skill for successful batting as the bowling length will often determine
101 the height at which the ball will arrive at the batter. Consequently, the time batters have to
102 adjust to post-bounce movement will also adjust depending on the bowling length – the balls
103 pitched further from the batter will deviate post-bounce earlier than those pitched closer to
104 batter (Leamon & Jones, 2021; Renshaw & Fairweather, 2000). By affecting the height at
105 which the ball arrives, bowling length could also determine the types of strokes that a batter
106 can play for a specific ball, which logically makes it an important determinant in deciding
107 whether a front of back-foot stroke is optimal. Previous research has revealed that some ball-
108 pitching lengths allow for limited stroke types, for example the “yorker” (which traditionally
109 pitches near the batter’s feet) is generally regarded as the hardest for a batter to strike (Moore
110 et al., 2012; Jamil et al., 2023), whereas other lengths could be met with multiple strokes, for
111 multiple outcomes (Leamon & Jones, 2021; Renshaw & Fairweather, 2000). This notion has
112 been reinforced by a recent study by Jamil et al. (2022) where certain bowling lengths (one full
113 and one short) and bowling lines (outside off stump) were revealed to be significantly
114 associated with the direction of power-hitting strokes that resulted in a maximum of 6-runs
115 being scored. Furthermore, stroke variety has also been related to bowling length, which shows
116 that balls bowled very close or very far from the batter often offer only one or two stroke
117 solutions whereas areas in between these zones allow for multiple strokes, often called
118 monostable and metastable zones respectively (Pinder et al., 2012). While this has been studied
119 in a lab setup from a batter’s perspective with several control measures and junior cricketers,
120 the same has not been examined at the highest level of competitive cricket over thousands of
121 balls bowled. It is also critical to note that this dependence of stroke availability according to
122 the location of the ball bounce (bowling line and length) – something initiated by the bowler
123 for every new delivery in a cricket match, makes this a highly interactive event, thereby making

124 it even more interesting to study decision-making. Keeping this in mind, this paper examines
125 whether a relationship exists between bowling length and type of strokes exhibited by elite
126 batsmen across two international 50-over cricket competitions.

127 **Method**

128 **Data and Sample**

129 The study of a relationship between bowling lengths and batter stroke types is divided
130 into two parts. First, we test whether a front-foot or back-foot stroke is required for successful
131 batting in international cricket. This is conducted through chi-square tests between stroke type
132 and outcomes of batting success – boundaries (4 or 6 runs scored on that ball) and (the lack of)
133 dot balls (0 runs scored on that ball). This would help understand whether front-foot and back-
134 foot stroke selections contribute to batting success. Next, a binomial logistic regression is
135 conducted with the different bowling lengths as predictors and batting stroke type (front-foot
136 or back-foot) as the outcome variable. A pattern of distribution of stroke types would indicate
137 the existence of a relationship between bowling lengths and batter strokes. The required
138 notations for this study were obtained from Opta (London, UK) - high levels of reliability have
139 been previously reported (Jamil et al., 2021), and these notations were further cross-validated
140 with ball-by-ball commentary logs of corresponding cricket matches which are available on
141 the official database of the International Cricket Council (ICC). Notations of bowling length
142 were provided by the data supplier and include classifications developed from commonly used
143 categories of approximate bounce distance from the batters' stumps; full toss (does not
144 bounce), yorker (approximately 0-2 m), half-volley (approx 2-5 m), length ball (or good length,
145 approx 5-8 m), back of a length (approx 8-10 m), short length (approx 10+ m) and the half-
146 tracker (also approx 10 m, but the ball bounces below waist height as it approaches the batter).
147 Foot movements were provided for each ball through the batting stroke types that were
148 exhibited by batters in intercepting each ball (see Table 1 for a description of all batting

149 strokes). A minority of strokes (<10% of total strokes analysed) that were not discernible as
150 front- or back-foot strokes, or those that denoted no movement were excluded from the
151 analysis. Moreover, only the bowling types ‘fast’ and ‘medium pace’ were included in this
152 investigation, as spin bowlers formulate a different category of bowling types with different
153 effects of bowling lengths and heights, thereby being characterised by interactions that may be
154 starkly different from pace bowling and require a separate analysis (Leamon & Jones, 2021).
155 Finally, illegal balls were also excluded from further analysis. After elimination and data
156 cleaning, a total of 19, 587 balls bowled in 60 cricket matches (120 innings) from two one-day
157 international cricket tournaments held in England were considered for analysis (ICC
158 Champions Trophy 2017 and the ICC Cricket World Cup 2019).

159 ***Insert Table 1 here***

160 **Statistical Analysis**

161 In order to understand whether a foot-based decision was required for batting
162 performance, Chi-Squared (χ^2) tests of independence were conducted in order to determine
163 whether there was any association between the front and back-foot strokes and batting
164 outcomes of 0 runs scored, as well as boundaries (4 or 6 runs) scored. The data was largely
165 formulated by nominal notations for each ball. Each ball bowled was treated as an independent
166 event due to the numerous factors of variation exhibited ball-by-ball (i.e. the timing of the
167 delivery, the condition of the ball – as it deteriorates with each ball bowled, the xy landing
168 location/direction of the ball, bowling speeds, the level of bounce post pitching, the degrees of
169 lateral movement/trajectory, the xy location of fielders and even environmental factors such as
170 humidity, temperature and wind speed). Each ball bowled therefore contributed to one and only
171 one cell in each of the χ^2 tests conducted in this study and passed McHugh’s (2013) assumption
172 conditions. In the case of statistical significance ($\alpha = 0.05$), standardized residuals were

173 calculated to identify those cells that made the greatest contribution to the test statistic (Sharpe,
174 2015). Moreover, given the dichotomous outcome variable as well as the two foot-based
175 responses, a 2x2 Fisher's exact table was used to understand whether there was a significant
176 association between foot-based movement and batting performance indicators of dots and
177 boundaries. To account for the number of cells present in the contingency tables, Bonferroni
178 corrections were applied (Sharpe, 2015) resulting in the adjusted ($p = 0.0125$) with the
179 associated critical values of ± 2.50 . Cramer's V effect sizes were also calculated (McHugh,
180 2013) and interpreted with the widely used thresholds of $0.1 \leq weak < 0.3$, $0.3 \leq moderate <$
181 0.5 , and $strong \geq 0.5$ (Cohen, 1988).

182 To understand whether there is a relationship between the bowling lengths and foot
183 movements shown by batters, the bowling length categories were used as predictors with the
184 front-foot/back-foot dichotomous outcome being the dependent variable in a logistic
185 regression. Coefficients (and p-values) were used to assess the relative differences between the
186 different ball-pitching length zones to the front-foot (and back-foot) outcome. Data pre-
187 processing was performed with the front-foot stroke coded as 1 and back-foot coded as 0, thus
188 a positive coefficient (B) would indicate a front-foot response in the logistic regression. To test
189 for reliability and out-of-sample validity of the resulting models, the performance of the
190 regression models was evaluated using a 5-fold cross validation alongside the traditional
191 method of significance levels. All analyses were performed using custom scripts in Python 3.8
192 and H2O library.3.14.

193 **Results**

194 As indicated in Table 2, front-foot and back-foot strokes are revealed to have
195 statistically significant associations with weak effects for both batting performance indicators
196 of facing dot balls (scoring 0 runs on that ball) ($\chi^2(1) = 99.798, p < .001$, Cramer's V = .071)

197 and scoring boundaries (4 or 6 runs) ($\chi^2(1) = 9.111, p = .003$, Cramer's $V = .022$). Although
198 the effect sizes are weak, this is to be expected as cricket performance is determined by various
199 technical, physical, tactical and contextual factors (Jamil et al., 2022). Post hoc analysis
200 through standardised residuals reveal that back-foot strokes present significantly greater than
201 expected dot balls (SR = 5.2), while the opposite is observed for front – foot strokes and dot
202 balls (SR= -5.0). For the performance measures boundaries, however, the standardised residual
203 does not cross the critical value for the adjusted $p = 0.0125$, and are therefore not considered
204 statistically significant.

205 After assessing whether front-foot and back-foot strokes are related to successful
206 batting performance outcomes, the results of the logistic regression for the front-foot/back-foot
207 outcome were used to understand whether there is an influence of bowling length on these
208 batter foot movements (front-foot and back-foot strokes). Results presented in Table 3 show
209 that all lengths except the half-tracker length ($p = 0.64$) showed a significant contribution in
210 correctly classifying strokes as front-foot or back-foot. That is, full toss ($B = 5.10, p < 0.01$),
211 yorker ($B = 4.12, p < 0.01$), half-volley ($B = 5.18, p < 0.01$) and length ball ($B = 3.71, p <$
212 0.01), all significantly predict predominance of front-foot stroke, while back of a length
213 (intercept) ($B = -2.31, p < 0.01$) and short length ($B = -0.93, p < 0.01$) significantly predict a
214 back-foot stroke. This indicates a clear relationship between bowling lengths and foot-based
215 stroke movements, specifically showing front-foot strokes emerging in balls bouncing
216 relatively closer to the batter (approximately up to 8 m) and back-foot strokes emerging in
217 response to balls bouncing further from the batter (further than 8 m). There is also a distinct
218 switch in preferred stroke type between the good length and short of good length zones. The
219 confusion matrix in Table 4 shows per class error rates for predicting front (~79% accurate)
220 and back foot (~93%) accurate strokes. Finally, the cross validation confirms that model was
221 able to predict which foot the shot was played on based on the pitching length of the ball with

222 high accuracy and reliability as shown in the cross-validation score (accuracy = 0.86 +- 0.002
223 and AUC = 0.88 +-0.005). Given the magnitude of coefficients and associated p-values, it can
224 be stated that there is a clear influence of bowling length on batter's front-foot and back-foot
225 decisions.

226 *****Insert Table 2 here*****

227 *****Insert Table 3 here*****

228 *****Insert Table 4 here*****

229 **Discussion**

230 The aim of this paper was to examine whether a relationship between bowling lengths
231 and batter decisions of front- and back-foot was observed in elite One-Day International cricket
232 tournaments. Chi-squared test results were used to explore significant associations between
233 front-foot and back-foot strokes and batting outcomes of boundaries and dots to first assess
234 whether any relationship between batting outcomes and foot-based batting decisions existed.
235 A logistic regression for foot-based batting strokes with different bowling lengths as
236 independent variables was subsequently used to further determine if (and how) specific batter
237 movements are related to different bowling lengths. Results from the chi-squared distribution
238 revealed a statistically significant relationship between successful batting outcome of run-
239 scoring (conceding or avoiding dot balls) and foot-based movements. Post-hoc results of the
240 standardised residuals show values that exceed the threshold of significance (± 2.50) as seen
241 in Table 2, showing greater than expected dot balls on the back-foot, and fewer than expected
242 dot balls on the front-foot. These results could indicate that batters prefer to meet the ball on
243 the front-foot in limited overs cricket, something that has been observed as a general pattern
244 by professional analysts in the sport (Leamon & Jones, 2021). Bowlers, on the other hand, may

245 prefer pushing batters on the back-foot where batting outcomes exhibit greater than expected
246 dot balls and fewer than expected boundaries, which aids in limiting run-scoring. However, as
247 there are no bowling length related insights in the chi-square output, the logistic regression
248 results are required to further understand this. Moreover, the results also showed weak effect
249 sizes for both dots balls and boundaries, which could be related to the high sample size and
250 subsequent effect of this on the p-value and effect size calculations. This is also why the chi-
251 square test results do not sufficiently inform us about the relationship between batting strokes
252 and bowling lengths. These results only show the general connection between successful
253 batting and foot-based decisions, and require results from the logistic regression to understand
254 the relationship of different bowling lengths with specific foot-based movements.

255 As shown in Table 3, all lengths except one (half-tracker) show a statistically significant
256 effect on the outcome variable of front-foot or back-foot stroke. The non-significant result of
257 the half-tracker zone may be explained by the relatively low sample size (N=103, 0.5% of total
258 balls) of balls bowled in this zone as compared to all the other zones included in the study.
259 Furthermore, this result could be partially explained by batter's responding to deliveries on this
260 length with deliberate movements made to avoid any contact with the ball that do not constitute
261 either a forward or backward movement (i.e. ducking or swaying out of the way), particularly
262 as this length has been known to cause injury (Pardiwala et al., 2018).

263 With regards to the remaining length zones, the findings clarify that accurately gauging
264 bowling length is a critical measure in determining front-foot and back-foot batting strokes.
265 Furthermore, a look at the sign of coefficients (positive sign indicates front-foot dominance
266 and negative sign a back-foot dominance) indicates the structure of front-foot and back-foot
267 strokes: when facing pace bowlers, front-foot strokes dominate all bowling lengths near the
268 batter, up to the good length zone, after which all the lengths exhibit a negative coefficient

269 indicating back-foot strokes. These findings involving elite batters are similar to results of a
270 lab study with junior-level batters facing balls bowled on different bowling lengths (Pinder et
271 al., 2012). The clear organisation of balls bowled closer to the batter being hit on the front-foot
272 and those further from the batter being hit on the back-foot show a coordinated organisation of
273 decisions (and subsequent actions) based on specific constraints initiated by the bowler
274 (bowling length). Such coordinated or interpersonal decision-making has been observed in
275 combat sports as well as certain dyads of team sports at a training level in the past (Krabben et
276 al., 2019; Kimmel & Rogler, 2018; Travassos et al., 2011; Hristovski et al., 2006), but rarely
277 at the elite competitive level, specifically in team sports. Studies that look at interactive
278 decision-making often also mention functional variability being exhibited in metastable zones
279 – a perceptual-motor state where multiple solutions can be used to arrive at the same outcome,
280 often situated between two monostable zones. In cricket, Pinder and colleagues (2012) found
281 this to be around the 6.5-7.5 m from the stumps in a lab study with junior cricketers, where
282 both front-foot and back-foot strokes are equally likely to occur. While this was not replicated
283 in the current study (largely due to different notations of length as well as no control measures
284 for body-scaling), the switch from front-foot to back-foot stroke is observed generally around
285 the same zone of bowling length – the length ball zone (approx. 6-8 m away from the stumps).
286 This finding warrants further inquiry as specific body-scaled dyads performing at the
287 international cricket level might display varying levels of functional variability as exhibited in
288 different stroke types – thereby serving as rare evidence of dynamic, interactive and
289 functionally adaptive decision-making at the elite team sport level (Araújo et al., 2019; Davids
290 et al., 2013).

291 The dominance of front-foot stroke observed in this length ball zone, often called the
292 “good length” zone in cricket parlance, may also reflect the need of batters to get on the front-
293 foot to score more runs in limited overs cricket (Leamon & Jones, 2021). It would be highly

294 informative, in this light, to examine how batters strategically adapt to varying formats of
295 cricket. This study included only one-day international matches where each team bats for a
296 maximum of 300 (legal) balls, but other cricket formats include batting for a maximum of 120
297 (legal) balls (Twenty20 cricket) or batting for unlimited balls (Test cricket) (See Mehta & van
298 der Kamp, 2021 for more information on formats). It is possible that batting strategies differ in
299 different formats and may be visible in batting outcomes, foot movement but also other
300 measures reflected batter choices such as stroke control or quality of contact. These choice-
301 driven measures can be used to compare batting strategies as well as transference between
302 formats – some of which may require a greater attacking impetus due to lesser time, such as
303 Twenty20. Similarly, analysts in professional cricket settings may benefit from such analyses
304 being performed for different locations across the world as atmospheric conditions could
305 potentially impact batting performances (Irvine & Kennedy, 2017; Mehta et al., 2022). The
306 analysis of these interactive outcomes can be viewed from the side of batter as well as bowler
307 decision-making, and can be used to on multiple levels like inter-player match-ups within an
308 over, an innings, but also between formats and locations.

309 In today's highly crowded and condensed cricket calendar, analyses of player
310 interactions at multiple levels could help the growth of strategic performance insights to
311 develop interventions that help players excel against specific players, match situations, formats
312 and conditions. Practical applications of this study could range from comparing effectiveness
313 of certain batters or bowlers in various conditions around the world to guiding training in build-
314 up to facing a certain opponent. Furthermore, coaches have been known to train batters to
315 execute specific strategies when facing balls bowled at certain bowling lengths such as the
316 length ball (often termed the good length ball) where their batting choices should exhibit high
317 levels of control (Connor et al., 2020). Studies like this could help understand how such aims
318 can be achieved and further improved upon, such as developing potential specifications for

319 stroke execution that integrate attacking or defensive outcomes while maintaining a focus on
320 control. This study was not without its limitations, this dataset included two competitions
321 played in England, and therefore warrants a similar inquiry in other cricketing hotspots around
322 the world. It included only competitions of the One-Day International format, while suggesting
323 the extension of such research into T20 and Test formats, as it would be difficult to integrate
324 these formats within the scope of this paper. Furthermore, data specifications of the bowler and
325 batter's dominant hand, trajectory specifications such as side of wicket and height of release
326 were also not available for this study, and were also not part of the aims of this study which
327 only looked to study the relationship between foot-based batter decisions and bowling length.
328 These variables could provide valuable information and could be studied separately in future
329 research.

330 **Conclusion**

331 This study aimed to understand whether a relationship between bowling lengths and
332 foot-based batter decisions existed in elite One-Day international cricket. By examining 19,587
333 balls bowled in two one-day international cricket tournaments, it was found that there was a
334 clear association between the occurrence of either front-foot or back-foot strokes based on the
335 bowling length initiated by the bowler. Specifically, the results of this study suggest that batters
336 in ODI cricket prefer front-foot strokes in their quest to score runs, which could in turn inform
337 bowling strategies during matches. Furthermore, the coordinative interdependence of foot-
338 based batting decisions and different bowling lengths discovered in this study signifies that
339 batting decisions are partly based on the bowler and the characteristics of the ball bowled. The
340 study sets a foundation of studying interdependent actions at the elite level, and can be applied
341 in studying elite batting in greater depth through a perspective that recognizes the interactive
342 and co-dependant nature of successful batting at the elite level of performance.

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