

Running Head: CHILDREN'S OPTION GENERATION

A Developmental Perspective on Option Generation and Selection: Children Conform to the  
Predictions of the Take-the-First Heuristic

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Abstract

Little is known about how children generate options for taking action in real-life situations or how they select which action option to actually perform. In this paper, we explore the interplay between option generation and selection from a developmental perspective using sport as a testbed. In a longitudinal design with four measurement waves, we asked 6- to 13-year-olds ( $N = 73$ ) to generate and select action options in a soccer-related task. Children conformed to predictions of the Take-the-First heuristic: They generated only a few options in decreasing order of validity (i.e., better options were generated earlier) and selected the first options they had generated. Older children selected the first option generated more often than younger children and generated options faster. Longitudinal effects revealed that both age groups generated fewer options and faster across waves. Time limitation fostered fewer and higher quality options being generated and selected. Overall, our results highlight the importance of considering the predecisional process of option generation to deepen our understanding of developmental changes in decision strategy use. Future research directions and implications for children's real-life decision making are discussed.

*Keywords:* option generation, option selection, Take-the-First heuristic, decision making, cognitive development

28 A Developmental Perspective on Option Generation and Selection: Children Conform to the  
29 Predictions of the Take-the-First Heuristic

30 Imagine being a young, talented soccer player. You are running through the midfield  
31 toward the goal, dribbling past one opponent after another. You are now 20 m from the goal,  
32 facing the opposing defense rapidly closing on you. What could you do? Shoot at the goal  
33 from where you are? Or should you pass the ball to one of your teammates—maybe the one  
34 approaching from the left? Making good and quick decisions is essential in sports, as in many  
35 other domains (Raab & Gigerenzer, 2015). Most often in real life, before actually deciding  
36 what to do, one has to think about what *could* be done, generating and simulating alternative  
37 actions that could be taken and imagining how possible scenarios could be played out.

38 Little is known about how decision-making strategies develop across childhood, and  
39 even less—if anything—is known about how children generate action or decision options and  
40 select among them. In this paper, we explore for the first time the interplay between option  
41 generation and selection, crucial building blocks of decision making, from a developmental  
42 perspective, using sports as a testbed.

43 **The Developing Decision Maker**

44 Most decision-making studies have focused either on adults or on the aging decision  
45 maker (Horn, Pachur, & Mata, 2015; Mata et al., 2012; Mata, Schooler, & Rieskamp, 2007).  
46 *Developing* decision makers, that is, children, have rarely been studied, and therefore the  
47 development of decision-making abilities across childhood is still poorly understood  
48 (Klaczynski, 2001). Decision-making research with children has focused on predecisional  
49 information search (i.e., the information children spontaneously ask for; see Ruggeri &  
50 Katsikopoulos, 2013; Ruggeri, Olsson, & Katsikopoulos, 2015; or the information children  
51 select from a set of informational items; see Davidson, 1991, 1996; Gregan-Paxton &  
52 Roedder John, 1995) or has investigated cue-based decision strategies (Betsch, Lehmann,  
53 Lindow, Lang, & Schoemann, 2016; Horn, Ruggeri, & Pachur, 2016; Mata, von Helversen,

54 & Rieskamp, 2011). Previous studies found that younger children (7- to 9-year-olds),  
55 compared to older children (10- to 12-year-olds) and adults, tended to search for more  
56 *irrelevant* information (Davidson, 1991), preferred more information-intensive strategies  
57 (e.g., strategies that collect and integrate all the information available), and had a harder time  
58 focusing on one or a few most informative cues when making decisions (Mata et al., 2011).  
59 Along the same lines, a recent study by Betsch and colleagues (Betsch et al., 2016) showed  
60 that neither preschoolers' nor primary school children's search was guided by the  
61 informativeness of the given cues.

62       To our knowledge, *option generation*, that is, the process of generating alternative  
63 action or decision options from which to select, has never been studied in children before.  
64 How many options do children generate and consider before making a selection? How *good*  
65 are those generated options, and are they generated in a random fashion or is the generation  
66 process systematic? Children start at an early age to make decisions for which they need to  
67 consider alternative options: what food to buy at the school canteen, what game to play, what  
68 club or hobby to commit to, what way to walk to school. Understanding the way children  
69 come up with and select alternative actions or decision options can shed light on the  
70 development of their decision-making strategies. We consider the development of decision-  
71 making strategies from an ecological rationality perspective. Within this framework,  
72 strategies are not good or bad per se, but rather, their effectiveness depends on the cognitive  
73 abilities of the decision-making agent, as well as on the characteristics of the environment  
74 considered. Thus, when studying the developing decision maker it is crucial to consider “the  
75 individual and [his or her] particular stage of ontogenetic development” (Todd, Gigerenzer,  
76 & the ABC Research Group, 2012, p. 11), also because the developmental stage influences  
77 the effect a given environment has on a person's use of heuristics (Marasso, Laborde,  
78 Bardaglio, & Raab, 2014).

79 **Option Generation and the Take-the-First Heuristic**

80 A decision-making strategy usually consists of a search, a stop, and a decision rule,  
81 which together define how and how much information has to be collected before one can  
82 make a decision (Gigerenzer, Todd, & the ABC Research Group, 1999). However, most real-  
83 world situations require people to generate alternative options *before* making a decision,  
84 rather than selecting one from a set of predefined options offered by an experimenter (Payne,  
85 Bettmann, & Johnson, 1988). Option generation has previously been studied with adults and  
86 adolescents in sports (Johnson & Raab, 2003; Raab & Johnson, 2007). Indeed, because of its  
87 naturally occurring dynamics (e.g., decisions to be made under time pressure; many potential  
88 alternative actions to be considered), sports is the ideal domain to test whether people use  
89 fast-and-frugal heuristics, such as the Take-the-First (TTF) heuristic (Raab, 2012; Raab &  
90 Gigerenzer, 2015).

91 The TTF heuristic is a cognitive model that captures option generation and decision  
92 making in familiar yet ill-defined tasks (Johnson & Raab, 2003; Raab, 2012; Raab &  
93 Johnson, 2007). The building blocks of TTF are formally defined as follows: a *search rule*,  
94 which generates options in order of validity (i.e., better options generated earlier), so that  
95 subjectively better options are generated earlier; a *stop rule*, according to which the  
96 generation phase should stop after two or three options have been generated; and a *decision*  
97 *rule*, according to which people should select one of the initial options generated (Johnson &  
98 Raab, 2003). Following TTF, people would generate only a few options and select the first  
99 one generated, rather than exhaustively generating and processing all possible options.  
100 Because these options were generated in order of validity, the decision, although fast and  
101 frugal, would tend to be accurate. Empirical studies have shown that the performance of  
102 experienced handball (Johnson & Raab, 2003), basketball (Hepler & Feltz, 2012), and soccer  
103 (Belling, Suss, & Ward, 2015) players is quite accurately predicted by the TTF heuristic:

104 Players generated about two options (e.g., shoot at the goal or pass to a teammate) in order of  
105 validity and selected the first option generated as the final decision.

### 106 **Time-Limitation Effects on Option Generation and Decision Making**

107 According to the ecological rationality framework (Todd et al., 2012), no strategy is  
108 *always* optimal, because the efficiency of a strategy depends on the environmental structure.  
109 In this sense, people should be *adaptive* and modify their strategies depending on how  
110 effective they are in a given environment. In many real-life situations, as in sports, decisions  
111 have to be made under limited time, and adults have been shown to adapt to time limitation  
112 by using faster and simpler strategies (Ben Zur & Brenitz, 1981; Payne et al., 1988). Along  
113 the same lines, in a study with adult soccer players, Belling and colleagues (2015) found that  
114 time limitation reduced the number of task-relevant options generated, although it did not  
115 impact the quality of players' decisions.

116 What about the effects of time limitation on the performance of developing decision  
117 makers? We know that children are *ecological learners*—they adapt their learning strategies  
118 to the characteristics (e.g., the statistical structure) of the task at hand (Horn et al., 2016;  
119 Nelson, Divjak, Gudmundsdottir, Martignon, & Meder, 2014; Ruggeri & Lombrozo, 2015),  
120 and they do so already by age 4 years (Ruggeri, Sim, & Xu, 2017). However, Davidson  
121 (1996) investigated the influence of time limitation on children's (7- to 10-year-olds)  
122 information search behavior and found that time pressure promoted faster, but generally not  
123 more selective searching.

### 124 **The Present Study**

125 In the present study we examined the development of children's option generation and  
126 selection by testing 6- to 13-year-old soccer players. In particular, we investigated whether  
127 children's option generation (search and stop rules) conformed to the predictions of the TTF  
128 heuristic. Additionally we tested the decision rule of TTF against other decision models: the  
129 random selection model, where the action to perform is chosen randomly from the set of

130 generated options; the Take-the-Best-Option heuristic, which predicts that children will select  
131 the best option (i.e., the option with the highest quality) among those generated; and the  
132 Take-the-Last heuristic, which predicts the selection of the last generated option. As children  
133 have been shown to use simple, noncompensatory information-search strategies (Bereby-  
134 Meyer, Assor, & Katz, 2004; Ruggeri & Katsikopoulos, 2013) and adolescent handball  
135 players have been shown to act according to TTF (Johnson & Raab, 2003), we expected  
136 children to make use of the TTF heuristic in a familiar real-life task. Taking into account  
137 previous developmental studies showing an increase in selective, noncompensatory strategy  
138 use with age (Davidson, 1991, 1996; Mata et al., 2011), we also expected older children to be  
139 more likely to conform to the predictions of TTF compared to younger children.

140       Whereas previous research has mainly used cross-sectional designs, in the present  
141 study we implemented a longitudinal design similar to that of Raab and Johnson (2007) that  
142 allowed us to monitor strategy change over time. We expected children to increase their  
143 reliance on fast-and-frugal heuristics across waves as they gained more experience with the  
144 task (cf. Raab & Johnson, 2007). More precisely, with a focus on the individual building  
145 blocks of TTF, we predicted that children would generate options faster (search rule; Raab &  
146 Johnson, 2007) and would generate fewer options (stop rule) across waves. Whether children  
147 would select the first option as their final choice more often across waves (decision rule) is  
148 more difficult to predict: Although theoretically an increase in experience should lead to  
149 selecting the first option more often as the final choice (Johnson & Raab, 2003; Raab &  
150 Johnson, 2007), no changes were found in the longitudinal study with adolescents (Raab &  
151 Johnson, 2007). Moreover, considering the general information-search literature that shows  
152 an increase in both a tendency to ignore irrelevant information and a selective focus on more  
153 informative cues across childhood (Davidson, 1991; Gregan-Paxton & Roedder John, 1995;  
154 Mata et al., 2011), we expected children to generate and select higher quality options across  
155 waves.



156 Finally, we explored whether and how time limitation influences children's option  
157 generation and selection. From the literature reviewed above it is unclear whether and how  
158 children would adapt their option generation and selection depending on the time available.

## 159 Method

### 160 Participants

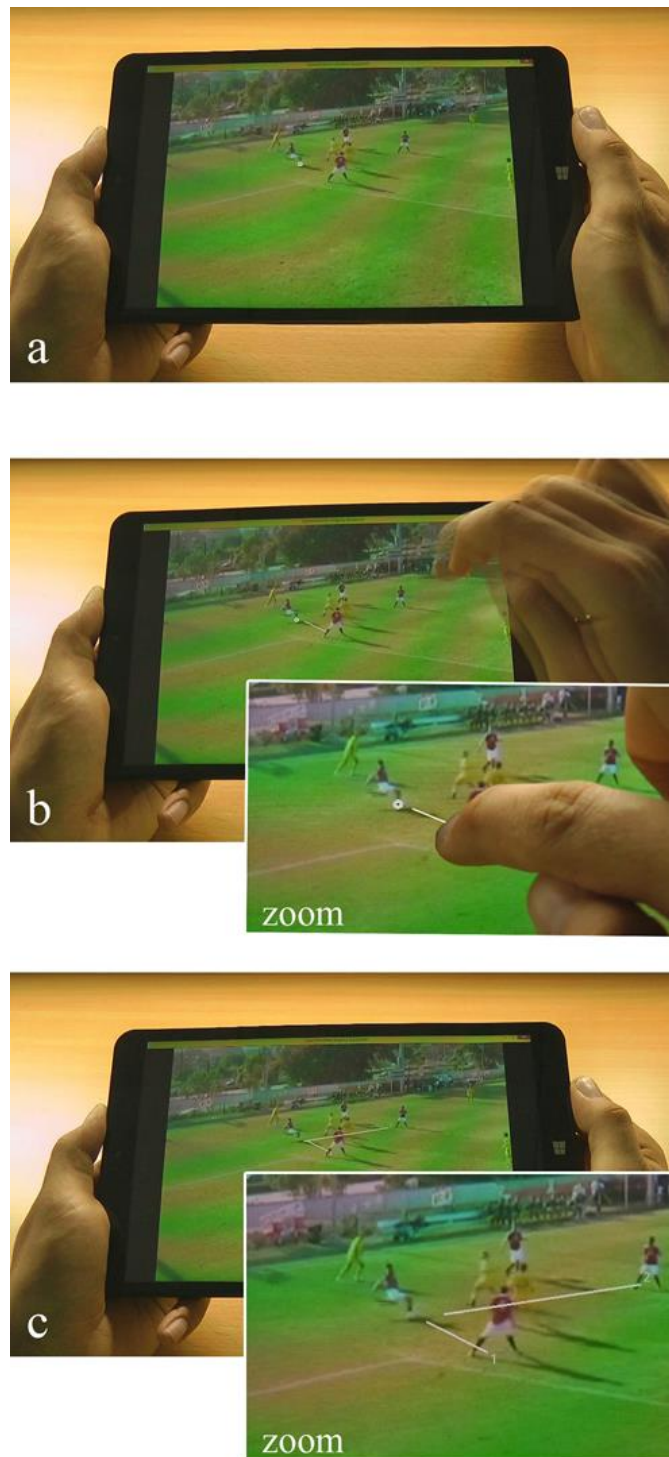
161 A total of 98 boys, recruited from a professional soccer academy in XXXXX,  
162 participated in this study. Using G-Power sample size estimation (Faul, Erdfelder, Buchner,  
163 & Lang, 2009), we estimated needing a sample of 66 participants ( $\alpha = .05$ ,  $1-\beta = 0.80$ ,  $f =$   
164  $0.42$  in the study of Belling et al., 2015). We recruited 98 participants to account for an  
165 expected dropout rate of about 25% across waves (cf. longitudinal study by Raab & Johnson,  
166 2007). Of the original sample, 73 completed all four measurement waves and were  
167 consequently included in the analyses: 38 younger children belonging to the Under-11 teams  
168 ( $M = 8.73$  years;  $SD = 1.15$  years; range = 6.67 to 10.50 years) and 35 older children  
169 belonging to the Under-14 teams ( $M = 12.37$  years;  $SD = 0.81$  years; range = 10.92 to 13.50  
170 years).

171 Most children ( $n = 65$ , 90%) were XXXXX; all children were XXXXX speaking and  
172 lived in or near a large city in western XXXXX. Before the start of the study, written  
173 informed consent was obtained from participants' parents and the local ethical review board  
174 approved the study protocol (XXXXXXXXXXXX).

### 175 Materials

176 We used 21 video scenes of live soccer match footage (three for the practice trials, 18  
177 for the test trials). We adopted the same task and materials as in Belling et al. (2015): After a  
178 short display of buildup play, the scenes suddenly stopped with a frozen frame, right before  
179 the player in possession of the ball had to make a decision (see Figure 1). Materials were  
180 presented to children on an 8.9" tablet.

181



182

183 *Figure 1.* Option-generation and selection procedure. (a) After a short display of buildup  
184 play, the scene stopped with a frozen frame, right before the player in possession of the ball  
185 had to decide which action to take. (b) Children generated alternative actions the player in  
186 possession of the ball could take by drawing them on the screen. (c) Children reviewed their  
187 generated options and selected the one they thought was the best.

188

## 189 **Design and Procedure**

190 We conducted the present study in a longitudinal cohort design (Schaie & Baltes,  
191 1975), in which two age groups of children were tested in four waves at intervals of 6 months  
192 (referred to as t1–t4; Wave 1: August 2015, Wave 2: February 2016, Wave 3: August 2016,  
193 Wave 4: February 2017). Overall, the study included three factors: measurement wave (four  
194 levels: t1–t4) and time limitation (two levels: short- or long-time condition) as within-subject  
195 factors, and age group (two levels: younger or older children) as between-subjects factor,  
196 resulting in a  $4 \times 2 \times 2$  design.

197 The task was administered to groups of five to nine same-aged children in a quiet  
198 room located at the soccer academy. Children, sitting alone at individual desks where a tablet  
199 was positioned, were introduced to the task procedure via a standardized instructional video  
200 (duration: 2:51 min) that was meant to familiarize them with the tablet and the task by  
201 walking them through the testing procedure. The experimental session consisted of 21 trials:  
202 The first three were practice trials, where children could ask the experimenter to clarify any  
203 questions. Only the results of the 18 test trials were included in the analyses. Each trial  
204 comprised two phases: option generation and option selection.

205 **Option generation.** On each trial, children were presented with a video of buildup  
206 play that stopped and held on a frame (see Figure 1 and Materials presented above). Children  
207 were then asked to generate a maximum of six action options (e.g., pass to the player on the  
208 right; dribble; shoot) directly marking them on the field using the touch screen (see Figure 1a  
209 and b). Trials were randomly assigned to either the short-time (9 trials) or the long-time (9  
210 trials) condition. In the long-time trials children were given 30 s to generate options, whereas  
211 in the short-time trials they were given 7.5 s to generate options. The order of presentation of  
212 the test trials was randomized.

213 **Option selection.** Children were presented with the action options they had generated  
214 in the previous phase and were asked to select the best option among these (see Figure 1c).

**215 Coding**

216 To assess the quality of the options generated and selected, two experienced youth  
217 soccer coaches, blind to the experimental hypotheses, independently evaluated all the options  
218 the children had generated for the 18 test trials. Both coaches had a UEFA B-level coaching  
219 license and at least 10 years of experience coaching a youth soccer team. For each of the 18  
220 test trials, presented in random order, coaches were asked to rate the options on a 10-point scale  
221 (from 1, 'not at all good', to 10, 'very good'). Having obtained good interrater agreement for  
222 the best option (Krippendorff's Kappa = .82,  $p = .01$ , intraclass correlation coefficient [ICC] =  
223 .77,  $p < .001$ ) and quality of all options generated ( $r = .56$ ,  $p = .01$ , ICC = .67,  $p < .001$ ), we  
224 computed the quality scores for each generated option by averaging coaches' quality ratings.

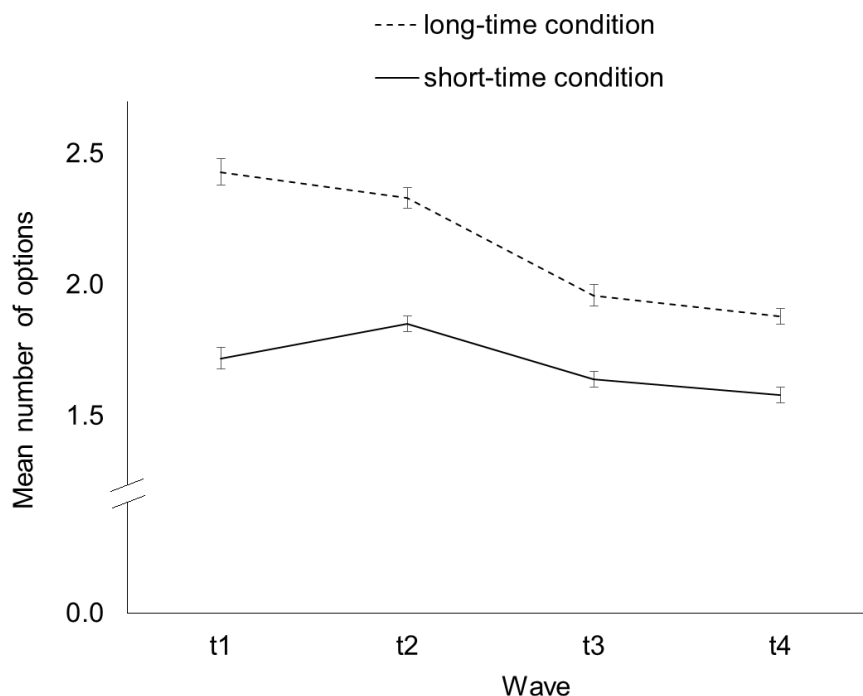
**225 Results**

226 First, we performed separate linear mixed-models analyses to investigate the effects  
227 of age group (two levels: younger vs. older children) as a between-subjects variable and wave  
228 (four levels: t1, t2, t3, t4) and time limitation (two levels: short-time vs. long-time) as within-  
229 subjects variables on four outcomes: (1) mean number of options generated across the 18 test  
230 trials; (2) average time taken to generate the first option; (3) average quality across all the  
231 generated options; and (4) average quality across all the selected options. Second, we  
232 interpreted the results in light of the predictions of the TTF heuristic (see above), further  
233 comparing them against predictions of the random selection model, the Take-the-Best-Option  
234 heuristic, and the Take-the-Last heuristic.

235 **Option Generation**

236 **Number of options generated.** Overall, in line with TTF, children stopped their  
237 generation after a mean of two options (1.92 options,  $SD = 0.99$ ). In 41.3% ( $n = 2,125$ ) of all  
238 trials, exactly two options were generated and in 35% ( $n = 1,822$ ) of all trials, only one option  
239 was generated. Older and younger children did not differ in the number of trials in which they  
240 generated exactly two options (younger children: 33.6%; older children: 49.5%;  $p = .081$ ).  
241 However, a chi-square test showed that older children generated only one option in fewer  
242 trials (24%) compared to younger children (45.7%),  $\chi^2(1) = 6.47$ ,  $p = .011$ , Cramér's  $V =$   
243 0.30. Also, in 2.1% ( $n = 111$ ) of all trials no options were generated. Older and younger  
244 children did not differ in the number of trials for which they generated no options (younger  
245 children: 1.4%; older children: 0.7%;  $p = .629$ ).

246 Our analysis revealed no effect of age group ( $p = .583$ ), but we did find main effects  
247 of wave ( $B = -0.22$ ,  $p < .001$ ) and time limitation ( $B = -0.75$ ,  $p < .001$ ) on the number of  
248 options generated, as well as a Wave  $\times$  Time Limitation interaction ( $B = 0.14$ ,  $p < .001$ ). In  
249 particular, the analysis showed that fewer options were generated across waves ( $M_{t1} = 2.08$ ,  
250  $SD = 1.19$ ;  $M_{t2} = 2.09$ ,  $SD = 1.00$ ;  $M_{t3} = 1.80$ ,  $SD = 0.86$ ;  $M_{t4} = 1.73$ ,  $SD = 0.80$ ) and that in  
251 the short-time condition children generated fewer options ( $M_{short} = 1.70$ ,  $SD = 0.84$ ) than in  
252 the long-time condition ( $M_{long} = 2.15$ ,  $SD = 1.07$ ). Moreover, the interaction effect revealed  
253 that in the long-time condition the number of options generated decreased across waves more  
254 dramatically than in the short-time condition,  $t(1195) = 9.44$ ,  $p < .001$ ,  $d = 0.52$  (see Figure  
255 2).



256

257 *Figure 2.* Number of options generated across waves (t1–t4) in the long-time and short-time  
 258 conditions. Error bars represent one SEM in each direction.

259

260 **Generation time of the first option generated.** The mean generation time of the first  
 261 option was 741.18 ms ( $SD = 386.11$  ms). All fixed factors—age group ( $B = 87.48$ ,  $p = .024$ ),  
 262 wave ( $B = -42.6$ ,  $p < .001$ ), and time limitation ( $B = -97.59$ ,  $p < .001$ )—influenced the  
 263 generation time of the first option. Older children ( $M_{\text{older}} = 691.70$  ms,  $SD = 351.91$  ms)  
 264 generated the first option faster than younger children ( $M_{\text{younger}} = 786.96$  ms,  $SD = 410.10$   
 265 ms). Options were generated faster across waves ( $M_{t1} = 827.29$  ms,  $SD = 446.09$  ms;  $M_{t2} =$   
 266  $735.36$  ms,  $SD = 378.99$  ms;  $M_{t3} = 703.12$  ms,  $SD = 360.48$  ms;  $M_{t4} = 700.19$  ms,  $SD =$   
 267  $338.54$  ms) and in the short-time condition ( $M_{\text{short}} = 689.68$  ms,  $SD = 339.75$ ;  $M_{\text{long}} = 790.70$   
 268 ms,  $SD = 420.24$  ms). No interactions between the fixed factors were apparent.

269           **Quality of the generated options.** The mean quality across all generated options was  
270 4.62 ( $SD = 2.79$ ). The analysis revealed no effect of age group ( $B = -0.14, p = .623$ ) or wave  
271 ( $B = 0.05, p = .468$ ) but did reveal a main effect of time limitation. The quality of all options  
272 generated was higher in the short-time condition ( $M_{\text{short}} = 5.26, SD = 2.79$ ) than in the long-  
273 time condition ( $M_{\text{long}} = 4.00, SD = 2.65; B = 1.3, p < .001$ ).

274           The first option generated had a mean quality of 5.20 ( $SD = 3.48$ ). The quality of the  
275 first option generated was not affected by age group ( $p = .951$ ) or wave ( $p = .328$ ) but was  
276 affected by time limitation ( $B = 1.00, p < .001$ ). Overall, children generated options of higher  
277 quality in the short-time ( $M_{\text{short}} = 5.71, SD = 3.36$ ) compared to the long-time ( $M_{\text{long}} = 4.71,$   
278  $SD = 3.53$ ) condition.

279           As predicted by TTF, children generated options in order of validity, which was  
280 confirmed by a repeated measures analysis of variance (ANOVA). The quality of the first  
281 three options generated differed significantly across serial positions<sup>1</sup>, Greenhouse–Geisser  
282  $F(1.46, 361.29) = 188.33, p < .001, \eta_p^2 = .43$ : The first options generated were of higher  
283 quality ( $M = 5.23, SD = 0.93$ ) compared to the second ( $M = 3.60, SD = 1.21$ ),  $F(1, 248) =$   
284  $401.96, p < .001, \eta_p^2 = .62$ , and third options ( $M = 2.83, SD = 2.07$ ),  $F(1, 248) = 315.33, p <$   
285  $.001, \eta_p^2 = .56$ . Children of both age groups generated options in order of validity as no age  
286 differences were apparent when considering the interaction with age group ( $p = .557$ ). The  
287 same pattern of results was also apparent when each wave was analyzed separately (please  
288 refer to the section S1 of the supplemental materials for the results reported by wave).

289           Our additional analysis revealed that the more options children generated, the less  
290 often their first option generated was the best of all their options,  $\chi^2(4) = 317.84, p < .001$ ,  
291 Cramér's  $V = .31$ . While children's first option generated was the best in 27.6% of the trials  
292 in which two options were generated, this was the case in only 3.4% and 0.5% for three and

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<sup>1</sup> We considered only those trials in which up to three options were generated (93%) to avoid the problem of too many missing points invalidating the results of the ANOVA.

293 four options generated, respectively. When five or six options were generated, the first option  
294 selected was never the best. The same trend was apparent for both, the younger ( $\chi^2(4) =$   
295  $115.87, p < .001$ , Cramér's  $V = .28$ ) and the older age group ( $\chi^2(4) = 199.57, p < .001$ ,  
296 Cramér's  $V = .33$ ).

## 297 **Option Selection**

298 **Quality of the selected option.** The mean quality of the options selected across trials  
299 was 5.00 ( $SD = 3.56$ ). Our analysis revealed no main effects of age group ( $p = .592$ ) or wave  
300 ( $p = .231$ ) on the quality of the final option selected. However, we found a main effect of  
301 time limitation ( $B = 0.79, p < .001$ ): Children selected options of higher quality in the short-  
302 time ( $M_{\text{short}} = 5.39, SD = 3.51$ ) compared to the long-time ( $M_{\text{long}} = 4.60, SD = 3.56$ ) condition.

303 **First option generated selected as final option.** Overall, children selected the first  
304 option they had generated as their final option in 75.9% of all trials and in 62.7% of trials in  
305 which more than one option was generated. Children selected options they had generated at  
306 earlier serial positions, particularly their first option generated, more often compared to  
307 options generated later in the generation phase (for all trials: all Cramér's  $V > .68$ ; for trials  
308 with more than one option generated: all Cramér's  $V > .59$ ). Generally, as predicted by the  
309 TTF decision rule, children selected the first option generated in more than 50% of the trials  
310 (for all trials: all Cramér's  $V > .43$ ; for trials with more than one option generated: all  
311 Cramér's  $V > .22$ ) and did so less often, the more options they generated ( $r < -.38$ , all  $p <$   
312  $.001$ ; see Table 1).

313 Considering only the trials in which more than one option was generated, neither  
314 wave ( $p = .770$ ) nor time limitation ( $p = .694$ ) had a significant impact on whether children  
315 selected the first as final option, but age group did ( $OR = 0.6, p < .001$ ). Older children ( $M_{\text{older}}$   
316  $= 67\%, SD = 47\%$ ) selected the first as final option significantly more often compared to  
317 younger children ( $M_{\text{younger}} = 57\%, SD = 50\%$ ).



318 Table 1

319 *Absolute Frequency of Selected Options Displayed by Serial Position and Number of*320 *Generated Options*

Number of generated options	Serial position of the selected option						Total
	1	2	3	4	5	6	
1	1,822	0	0	0	0	0	1,822
2	1,461	664	0	0	0	0	2,125
3	472	223	190	0	0	0	885
4	110	31	27	45	0	0	213
5	26	11	9	8	8	0	62
6	14	7	3	4	2	8	38
Total $n_{\text{all trials}}$	3,905	936	229	57	10	8	5,145
Total % $_{\text{all trials}}$	75.9%	18.2%	4.5%	1.1%	0.2%	0.2%	100.0%
Total $n_{\text{trials in which more than one option was generated}}$	2,083	936	229	57	10	8	3,323
Total % $_{\text{trials in which more than one option was generated}}$	62.7%	28.2%	6.9%	1.7%	0.3%	0.2%	100.0%

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322 **Model comparison.** Considering only those trials in which more than one option was  
323 generated, children selected the best (i.e., highest quality) among the generated options  
324 (Take-the-Best-Option heuristic) in 24.4% of the trials. In 18.6% of the trials, taking the best  
325 option meant following the TTF decision rule; in 5.8% of the trials, children selected the best  
326 but not the first among their options generated, and in 44.1% of the trials, they selected the  
327 first but not the best option. Children selected their *last* option in 27.5% in trials. Selection of  
328 the last option never corresponded to the TTF decision, by definition.

329 Overall, children selected the first option more often compared to what was predicted  
330 by the random selection model,  $t(3322) = 23.78, p < .001, d = 0.41$ ; the Take-the-Best-Option  
331 model (24.4%),  $\chi^2(1) = 559.08, p = .003$ , Cramér's  $V = .43$ ; and the Take-the-Last model  
332 (27.5%),  $\chi^2(1) = 455.04, p = .001$ , Cramér's  $V = .39$ . Please refer to Tables S1.2 and S1.3 in  
333 the supplemental materials for results of the model comparison reported by wave.

334 In an additional exploratory analysis, we tested whether an increasing number of  
335 options generated decreased the likelihood of selecting the first, best, and last option. Results  
336 showed that the more options children generated, the less often they selected their first ( $\chi^2(4)$   
337 = 99.90,  $p < .001$ , Cramér's  $V = .17$ ), best ( $\chi^2(4) = 452.40$ ,  $p < .001$ , Cramér's  $V = .37$ ), and  
338 last option ( $\chi^2(4) = 42.83$ ,  $p < .001$ , Cramér's  $V = .11$ ). The same pattern emerged for both  
339 age groups. Irrespective of the number of options generated, older children selected the first  
340 option generated when it was the best one more often (21.4%) than younger children  
341 (15.4%),  $\chi^2(1) = 17.50$ ,  $p < .001$ , Cramér's  $V = .07$ .

## 342 Discussion

343 Little is known about how children generate and select options for taking action in  
344 real-life situations. In this paper we explored the interplay of option generation and selection,  
345 crucial building blocks of decision making, from a developmental perspective, testing  
346 children in a sport-specific task. In particular, taking an ecological rationality perspective, we  
347 tested whether the TTF heuristic could predict children's option generation and selection  
348 better than other cognitive models.

### 349 Children Use the TTF Heuristic

350 Our results showed that children's option generation and selection generally  
351 conformed to the predictions of the TTF heuristic: They generated on average about two  
352 options per trial and generated them in a meaningful way, that is, producing higher quality  
353 options first. That children did apply the TTF heuristic in a real-life decision-taking task is  
354 consistent with findings showing that even school-aged children use decision heuristics that  
355 match the task at hand (e.g., Horn et al., 2016) and results demonstrating children's use of  
356 simple, noncompensatory information-search strategies (Bereby-Meyer et al., 2004; Ruggeri  
357 & Katsikopoulos, 2013).

358 Children's option generation influenced their final selection: For both younger and  
359 older children, the more options they generated, the less often they selected the first option.

360 This pattern, that is, the mismatch between the first option generated and the one selected,  
361 has been referred to as *dynamic inconsistency* and has been shown to increase with the  
362 number of options generated (Johnson & Raab, 2003; Raab & Johnson, 2007). Thus, our  
363 results indicate that the decision rule children apply depends, at least to some degree, on their  
364 stop rule, such that children's decisions are more dynamically inconsistent when they stop  
365 later, after having generated more options. Recent research has identified the stop rule as a  
366 crucial factor responsible for younger children's general lower efficiency in information  
367 search compared to that of adults (Ruggeri, Lombrozo, Griffiths, & Xu, 2016). On the same  
368 line, in the present study children were more efficient when they had generated fewer  
369 options: The more options younger and older children generated, the less likely they were to  
370 select the first or the best option. Importantly, children's first option selected was also less  
371 likely to be the best the more options they had generated, which was true for younger and  
372 older children alike.

373 That children do indeed use the TTF heuristic was further supported by our model  
374 comparisons: Children's selection was more consistent with the predictions of TTF,  
375 compared to the random, Take-the-Best-Option, or Take-the-Last models. Importantly,  
376 children selected the first option in most of the decisions made.

377 Although the number and quality of options generated did not differ between age  
378 groups, older children generated options faster. As hypothesized, older children selected the  
379 first option generated more often than younger children. These results can be interpreted as  
380 an indication of older children having a stronger and more selective decision rule and are in  
381 agreement with previous findings showing that preschoolers and elementary school children  
382 are not yet able to selectively attend to the most relevant information (Betsch et al., 2016;  
383 Mata et al., 2011). The results further document a shift to a more pronounced use of  
384 noncompensatory strategies by the age of 11 years (Mata et al., 2011). Importantly, our  
385 results underline that following the simple decision rule by "taking the first" did not always

386 yield to selecting of the best option. Indeed, selecting the first option did not lead children to  
387 select the best option in many (44.1%) of the trials. Finally, although no age differences  
388 emerged for the quality of the options generated or selected, we observed that older children  
389 selected their first option generated when it was the best one more often (21.4%) than  
390 younger children (15.4%). In this sense, our results suggest that older children's option  
391 generation and selection strategies are more effective than those of younger children.

### 392 **Longitudinal Effects on Option Generation**

393         Like the adolescent handball players in the study of Raab and Johnson (2007),  
394 children of both age groups in the present study sped up their option generation and generated  
395 fewer options across the four measurement waves. However, the quality of the options  
396 generated and selected was not affected by wave. Contrary to our predictions, children did  
397 not select the first option generated more often across waves and, more generally, seemed not  
398 to modify their decision rule in the course of the 1.5-year testing period. This result can be  
399 interpreted in at least two different ways, not mutually exclusive. First, the gain in domain-  
400 specific experience across waves was not enough to shift the decision rule application (Horn  
401 et al., 2016; Raab & Johnson, 2007). In this sense, children's experience across waves might  
402 not have been enough for them to learn how to implement more effective selection strategies,  
403 also because no feedback was offered. Second, there might have been a ceiling effect:  
404 Because the children were already selecting the first option generated at a high percentage in  
405 the first measurement wave, the potential to increase their reliance on this decision rule  
406 across waves was limited.

### 407 **Time Limitation Fosters Better Options and Decisions**

408         In contrast with the results obtained with adult soccer players (Belling et al., 2015),  
409 when less time was available, children generated fewer options and selected options of higher  
410 quality. Indeed, in line with the notion of "less-is-more" and in theoretical agreement with the  
411 ecological rationality perspective (Johnson & Raab, 2003; Todd et al., 2012), the time



438 situations, such as traffic conditions, impact children's option generation and selection.  
439 Systematically manipulating environmental constraints across computer-based or real-life  
440 tasks will shed light on children's ability to adapt their decision-making strategies in real  
441 time. What is learned could inform the development of age-tailored interventions focusing on  
442 prevention (e.g., traffic education) and training (e.g., sports, physical education).

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