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## Search for transitive connections

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### Abstract

We present an algorithm that detects non transitive connections in the game of Go. An optimized Alpha–Beta search is used on top of two Generalized Threats Searches, one for each of the two connections. It deals with full board situations such as the ones encountered in real games. Our program is able to solve problems such as the double monkey jump or the double keima on the second line. Even if the results are not theoretically perfect, they are pretty reliable given the results on a test suite.

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### 1. Introduction

In this paper, we explore a way to reduce the complexity of a search on a double connection by searching both connections separately as much as possible. In the best case the two connections are independent, and a complex search that would cost  $O((2p)^{2d})$  can be reduced to two searches of complexity  $O(p^d)$ , the variable  $p$  being the average number of possible moves in a connection game, and  $d$  the depth of the search for solving one of the connection game.

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23 In practice, the two connections are seldom perfectly independent. Even when  
24 they are not independent, it is useful to search the two connections separately  
25 as it helps finding sets of relevant moves.

26 Programs have weaknesses at finding non transitivity as can be seen in  
27 tournament games [1]. Some strong Go programs handle non transitivity using  
28 hand-coded patterns. For example, Many Faces of Go uses a few hundred such  
29 patterns. Since there are too many cases of non transitivity, this approach is  
30 limited.

31 The Section 2 describes the problem of the non transitivity of connections.  
32 The Section 3 outlines the adaptation of Generalized Threats Search to the  
33 connection game. The Section 4 details the evaluation function and the selec-  
34 tion of moves used in our transitive connection search algorithm. The Section  
35 5 gives experimental results. Eventually, the Section 6 concludes and outlines  
36 future work.

## 37 2. Non-transitivity of connections

38 In this section we define the problem of the non transitivity of connections.  
39 Even the best computer Go programs have problems dealing with non transi-  
40 tivity. This problem is a special case of the more general problem of the depen-  
41 dence between two or more different sub-games.

42 Let  $A$ ,  $B$ ,  $C$  be three stones of the same color,  $C_{AB}$  the connection between  $A$   
43 and  $B$  and  $C_{BC}$  the connection between  $B$  and  $C$ . We call max player the player  
44 who wants to connect and min player the player who wants to disconnect. In  
45 the starting position, the connections  $C_{AB}$  and  $C_{BC}$  must be won; otherwise the  
46 transitive connection would not be won a fortiori.

47 The easiest case is when the connections  $C_{AB}$  and  $C_{BC}$  are independent.  
48 Then the transitive connection is won. Fig. 1 shows an example.

49 The interesting cases are when the connections are not independent, that is  
50 to say when there is at least a move by min player which threatens to break  
51 both connections. Then the transitive connection may or may not be won.

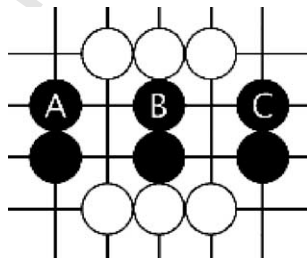


Fig. 1. Two independent connections.

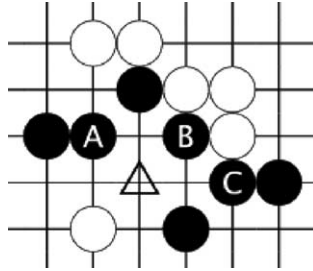


Fig. 2. Non transitive connections.

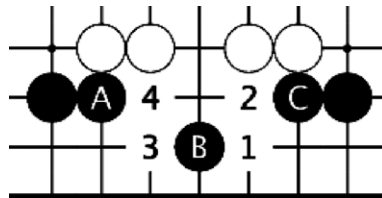


Fig. 3. Transitive connections.

52 Fig. 2 shows an example where the transitive connection is lost if min player  
 53 moves first: the move that disconnects is white  $\Delta$ . Now Fig. 3 shows an exam-  
 54 ple of two connections that are not independent but which make a transitive  
 55 connection nonetheless. The connections  $C_{AB}$  and  $C_{BC}$  are not independent be-  
 56 cause a white move at 1 threatens to break both; but then a black move at 2  
 57 would repair both connections.

### 58 3. Generalized Threats connection search

59 In order to find transitive connections, we have to solve the problem of find-  
 60 ing direct and single connections. This section is about the single connection  
 61 game. We have used Generalized Threats Search [2] to solve the single connec-  
 62 tion game. The threat used for these experiments is the (8, 5, 2, 0) general threat.  
 63 Moves in a Generalized Threats are associated to an order. The order of a po-  
 64 sition is the number of moves in a row by the same player that are required to  
 65 win the game. Moves of order  $n$  are moves associated to positions of order less  
 66 or equal to  $n$ . In the (8, 5, 2, 0) threat, eight is the number of order one moves  
 67 allowed in the Generalized Threat, five the number of order two moves and  
 68 two the number of order three moves. Only threats that have less moves for  
 69 each order are verified at min nodes.

70 It is not mandatory to use Generalized Threat Search; however, whatever  
 71 algorithm we use to find direct connections, it has to send back what we call

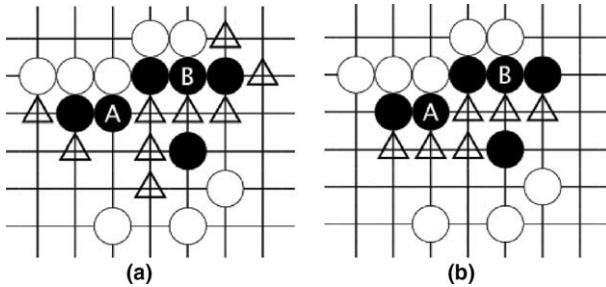


Fig. 4. Two possible traces for a connection.

72 a trace. Roughly speaking; the trace is a set of intersections that may change  
73 the result.

74 Precisely, the trace is a set of empty intersections such that if none is mod-  
75 ified, the result of the search associated to the trace is not modified. In order to  
76 compute the trace of a connection, we have applied the following principle: for  
77 each test in the search, add to the trace the intersections that enable the test to  
78 be true. In most cases, we have to choose some intersections among many to be  
79 included in the trace. For example if the test is that the string has at least two  
80 liberties, any pair of liberties could be added. In practice, there are different  
81 possible traces that can be associated to a given search. In Fig. 4 we can see  
82 the difference between the trace found by our program for a connection, and  
83 a minimal trace obtained by hand.

84 The evaluation function for the single connection returns:

- 85 • Lost if one of the two strings to connect is captured in a ladder,  
86 • lost if no max moves have been found, usually because the path between the  
87 two strings is too long and the common adjacent strings have too many  
88 liberties,  
89 • won if the two stones to connect are in the same string,  
90 • unknown otherwise.

91  
92 We use specialized functions to find the max moves. The order of a threat is  
93 the number of moves in a row the max player has to play in order to win the  
94 game [2]. The functions called for finding the max moves depend on the order  
95 of the threat. We have different specialized and heuristic functions for finding  
96 possible moves that connect in one move, in two moves or in three moves.  
97 When there is a possibility for one of the two strings to be captured, the only  
98 max moves considered are the moves that save the threatened string. Concern-  
99 ing the min moves, the Generalized Threat Search algorithm uses the trace of  
100 the verified threat to find the relevant min moves.

## 101 4. Search for transitivity

102 We use an Alpha–Beta algorithm with transposition tables, two killer moves  
103 and the history heuristic. The game specific functions of our Alpha–Beta are:  
104 the evaluation function, and two functions *minMoves* and *maxMoves*, which  
105 return sets of relevant moves for the players min or max.

### 106 4.1. Evaluation function

107 The evaluation function searches the connections  $C_{AB}$  and  $C_{BC}$  in isolation  
108 using the Generalized Threats Search algorithm, and tries to deduce from this  
109 the status of the transitive connection. The situation is different depending on  
110 who is to play.

111 We consider first the case where max player is to play. The connections  $C_{AB}$   
112 and  $C_{BC}$  are first searched with max player playing first, then with min player  
113 playing first. Besides the results, the searches also return the traces of all inter-  
114 sections on which the results depend. There are two cases where we can be sure  
115 of the status of the transitive connection:

- 116 • If  $C_{AB}$  or  $C_{BC}$  is lost, assuming max player plays first, then the transitive  
117 connection is lost.
- 118 • If one of the connections, say  $C_{AB}$ , is won (assuming min player plays first),  
119 if the other,  $C_{BC}$ , is winnable (i.e. it can be won if max player plays first), and  
120 if the traces on which those two results depend are disjoint, then the transi-  
121 tive connection is winnable. Indeed, in order to win it suffices for max player  
122 to play the winning move in connection  $C_{BC}$ .

123 We now consider the case where the player min (i.e. the player to discon-  
124 nect) plays first. There are again two cases where we can be sure of the status  
125 of the transitive connection:  
126

- 127 • If  $C_{AB}$  or  $C_{BC}$  is lost, assuming min player plays first, then the transitive con-  
128 nection is lost.
- 129 • If both connections  $C_{AB}$  and  $C_{BC}$  are won, assuming min player plays first,  
130 and if the traces on which those two results depend are disjoint, then the  
131 transitive connection is won.

132

### 133 4.2. Choose of min moves

134 In case the evaluation function cannot decide the status of the transitive  
135 connection we have to continue the main Alpha–Beta search which deals with  
136 the transitive connection as a whole. Hopefully the searches that have been

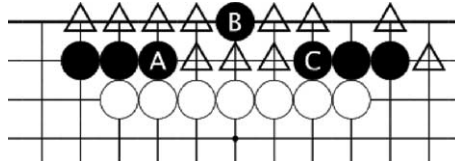


Fig. 5. Min moves.

137 made on the connections  $C_{AB}$  and  $C_{BC}$  can give valuable information to find a  
 138 relevant set of moves.

139 We take as set of min moves (moves for min player) the moves that threaten  
 140 to break either connection  $C_{AB}$  or  $C_{BC}$ . This is the union of the traces of the  
 141 two searches that have shown that the connections  $C_{AB}$  and  $C_{BC}$  are won when  
 142 min player plays first. An example of a set of relevant min moves found by our  
 143 program is given in Fig. 5.

144 It is possible to use the intersection of the traces rather than the union. This  
 145 is less safe as can be seen for instance in problem 13 of our test suite (Fig. 10),  
 146 where the only move that disconnects may not be in the intersection. In prac-  
 147 tice, using the intersection of the traces solves more problems and takes less  
 148 time as can be seen in the experimental results section.

149 In fact, even taking the union of the traces as the set of min moves is not  
 150 perfectly safe. We have found that it is safe for the problems of our test suite,  
 151 but we have built a transitivity problem (Fig. 6) where it misses a move. The  
 152 move white 1 does not threaten either connection  $C_{AB}$  or  $C_{BC}$ , but it does break  
 153 the transitive connection because black cannot defend against both white 2 and  
 154 white 3. One can note that white 3 would have directly worked too, so even in  
 155 this problem we would find at least one disconnecting move.

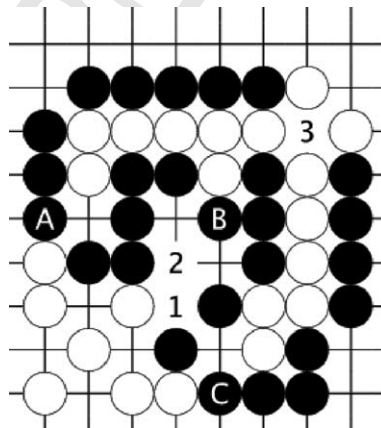


Fig. 6. Pathological position.

## 156 4.3. Choose of max moves

157 In order to find a set of max moves, it is not suitable to use traces as for min  
 158 moves; instead it is better to use the notion of *order*. The order of a connection  
 159 is the number of moves in a row that are needed to join the two strings in the  
 160 same string, if the opponent does not play [2]. Fig. 7(a) shows max moves to  
 161 connect of order 2. There is a path of length 2 composed of empty intersections  
 162 between strings *A* and *B*. Connecting may also involve capturing opponent  
 163 strings adjacent to both strings *A* and *B*. Fig. 7(b) gives an example of the order  
 164 2 connection moves related to capturing a common adjacent string. Fig. 7(c)  
 165 details the moves of order 3 to connect strings *A* and *B*.

166 Our algorithm to find order *n* moves between the strings *A* and *B* is shown in  
 167 Fig. 8. In case this algorithm is applied to a connection of order less than *n*, it  
 168 will not return all the moves of order *n* (which would be all the legal moves!),  
 169 only those close to the connection, which is usually an advantage.

170 The set of max moves depends on the order at which we want to search the  
 171 transitive connection. We have chosen to take as set of max moves the union of  
 172 the sets of moves of order up to *ordermax* in each connection  $C_{AB}$  and  $C_{BC}$ .  
 173 The variable *ordermax* equals at most the order of the maximum threat for  
 174 the connection search plus one.

175 An example of a set of max moves found by our program is given in Fig. 9.  
 176 In this case it is obviously far from perfect, because our program selects moves  
 177 of order 3 although the two connections are of order 2.

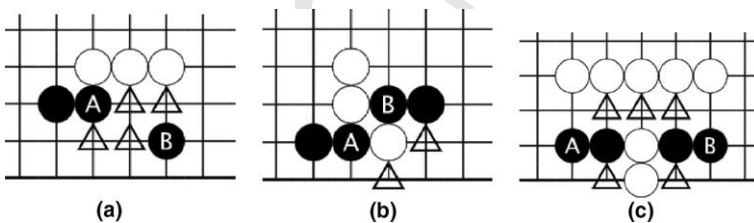


Fig. 7. Moves of order 2 (a and b) and 3 (c).

$$S \leftarrow \emptyset$$

For each move *m* at a liberty of string *A*, or at a liberty of an opponent string adjacent to *A* that has at most *n* liberties and which is either adjacent to *B*, or that has a common liberty with *B* :

$$\text{Play}(m)$$

If the connection is of order  $n - 1$  :

$$S \leftarrow S \cup \{m\} \cup \{\text{moves of order } n - 1\}$$

$$\text{Undo}(m)$$

return *S*

Fig. 8. Algorithm to find order *n* moves between strings *A* and *B*.

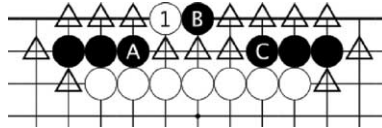


Fig. 9. Set of moves when Black is to play.

## 178 5. Experimental results

179 In our experiments, the maximum threat used for the Generalized Threats  
 180 connection search has been set to (8, 5, 2, 0). We only choose moves of order  
 181 less or equal to three for the connections  $C_{AB}$  and  $C_{BC}$  in order to find the max  
 182 moves in the transitive search. The maximum number of moves in each of the  
 183 two Generalized Threats Searches is limited to 100,000 unless stated otherwise.

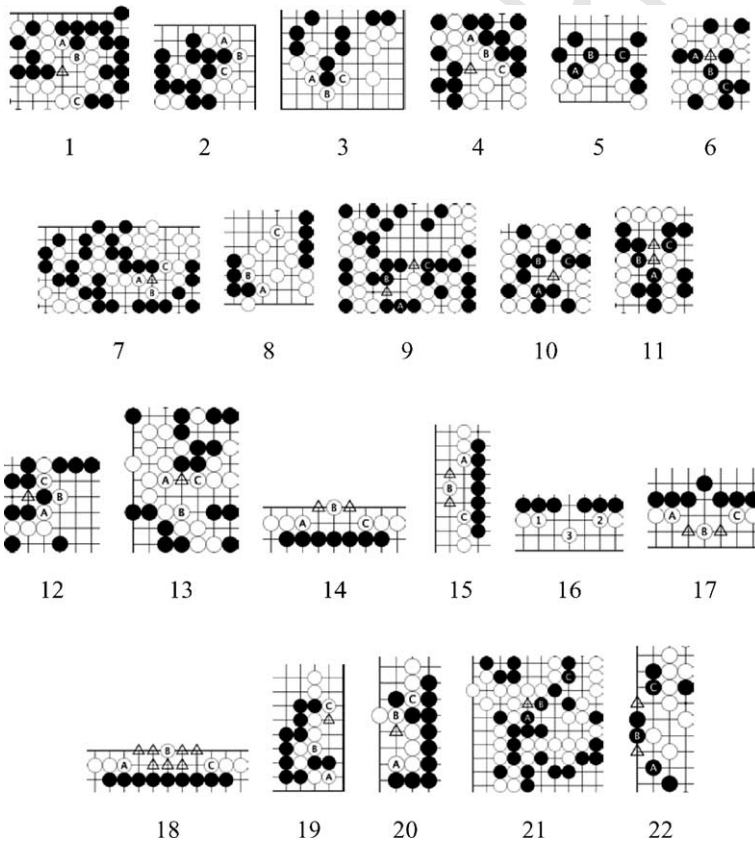


Fig. 10. Problems of the test suite.



184 We have built a test suite of 22 problems depicted in Fig. 10. The discon-  
 185 necting moves, if any, are marked. Only problem 17 involves a ko. Two thirds  
 186 of the problems are taken from Golois games, and one third are classical prob-  
 187 lems. The problems from Golois are taken from games where it failed to ana-  
 188 lyze transitivity correctly. This test suite is available for download on the first  
 189 author's web page.

190 In order to compare transitive search with another algorithm, we have tested  
 191 an optimized Alpha–Beta algorithm that uses Golois moves generators for con-  
 192 nections and disconnections. The evaluation function of this Alpha–Beta re-  
 193 turns Lost as soon as one of the two connections is Lost. A connection is  
 194 Lost if the length of any path between the two strings is strictly greater than  
 195 three. A connection is Won if the two stones to connect are in the same string.

196 In Table 1, for each problem, the number of moves played in the search and  
 197 the elapsed time used for the search are given. The experiments were run on a  
 198 3.0 GHz Pentium with 2 GB of RAM. The maximum number of nodes is set to  
 199 10,000,000 in the transitive search and in the Alpha–Beta. The left part of the

Table 1  
Nodes and time for the transitivity problems

Problem	Alpha–Beta			Transitive search (with union of the traces)		
	Moves	Time (ms)	Solved	Moves	Time (ms)	Solved
1	61,188	190	Yes	115,065	180	Yes
2	348,950	1,390	Yes	56,130	90	Yes
3	85,582	330	No	397,880	560	No
4	147	10	Yes	1,857	0	Yes
5	2,687,471	6,850	Yes	384,933	620	Yes
6	8,655	40	Yes	11,670	20	Yes
7	234,806	1,910	Yes	86,656	160	Yes
8	43,537	180	Yes	5,920	10	Yes
9	72,800	320	Yes	6,231	0	Yes
10	101	0	Yes	6,503	10	Yes
11	8,307	100	Yes	31,742	50	Yes
12	88,152	400	Yes	93,959	160	Yes
13	10,000,108	39,510	No	1,906,627	3,330	No
14	21,266	40	Yes	27,442	40	Yes
15	198,569	610	Yes	372,134	490	Yes
16	10,000,256	33,300	No	10,437,826	14,320	No
17	10,000,183	29,530	No	783,545	1160	No
18	3605	10	Yes	269,258	430	Yes
19	3,828,385	16,360	Yes	109,155	220	Yes
20	22,871	50	Yes	10,186,430	14,900	No
21	988	0	Yes	6,436	260	Yes
22	13,703	20	Yes	18,846	40	Yes
Total			18			17

200 table details the performance of the Alpha–Beta algorithm, and the right part  
201 the performance of the Transitive search algorithm.

202 Concerning the results of the Transitive search algorithm with union of the  
203 traces, problem 3 is not solved because in one of the forced lines a black cutting  
204 string gains 4 liberties and is therefore considered stable. In problem 17, our  
205 algorithm finds the good move but fails to see that it depends on a ko. In prob-  
206 lem 20, the program fails because it does not verify that connected strings are  
207 not captured in a ladder, and therefore the trace does not contain the capturing  
208 move for the attacker. In problem 16, it fails because it is short of nodes in the  
209 main search, and in problem 13 it fails because it is short of nodes in one of the  
210 connection searches. The problem 13 involves two opponent strings that can-  
211 not be captured, this is why our algorithm fails to see the disconnection: for  
212 each move in the first capture search, it searches the other capture. In order  
213 to solve it faster, we should decompose it into two independent capture prob-  
214 lems. This is not currently done by our connection algorithm.

215 **Table 2** details the runs of the Transitive search algorithm using the intersec-  
216 tion of the traces for min moves instead of the union as in **Table 1**. The algo-  
217 rithm using the intersection is faster, especially for problem 16 that is solved

Table 2

Transitive search with intersection of the traces

Problem	Moves	Time (ms)	Solved
1	104,025	150	Yes
2	2,558	10	Yes
3	359,609	490	No
4	1,502	0	Yes
5	5,962	10	Yes
6	10,463	10	Yes
7	23,268	50	Yes
8	1,230	0	Yes
9	6,231	10	Yes
10	5,471	0	Yes
11	14,262	20	Yes
12	93,959	130	Yes
13	1,913,120	2,720	No
14	19,165	20	Yes
15	194,050	250	Yes
16	1,214,948	1,660	Yes
17	382,636	490	No
18	67,642	90	Yes
19	99,858	150	Yes
20	145,989	250	No
21	4,298	10	Yes
22	7,723	10	Yes
Total			18

Table 3

Number of problems solved depending on max. time, algorithm and max. secondary nodes for Transitive search

Time (ms)	Transitive search					Alpha–Beta
	1000	5000	10,000	30,000	100,000	
10	7	6	9	9	8	4
30	7	10	10	11	11	5
100	7	13	14	13	13	9
300	7	13	15	15	17	11
1000	7	13	15	15	17	15

218 relatively quickly compared to Alpha–Beta and Transitive Search that do not  
 219 solve it because of a lack of nodes. In problems 5 and 8, connections are transi-  
 220 sive. These two problems are representative of the usual problems that a transi-  
 221 sive search algorithm has to solve (i.e. connections are usually transitive). If  
 222 we compare the transitive search algorithm using intersection of the traces with  
 223 the Alpha–Beta algorithm, we see that for these problems, the transitive search  
 224 algorithm is much faster.

225 The transitive search can be tuned with two parameters: the maximum transi-  
 226 tive search time and the maximum number of nodes allowed to each connec-  
 227 tion search. Connections searches are also named secondary searches. The  
 228 Transitive search algorithm with intersection of the traces is used for the exper-  
 229 iments. In order to choose the best algorithm for a given maximum response  
 230 time, we have tested Transitive search for different secondary nodes and differ-  
 231 ent maximum response time. Table 3 gives the number of problems solved  
 232 depending on the two parameters. When choosing the appropriate number  
 233 of secondary nodes for a given maximum time, we see that Transitive search  
 234 is better than Alpha–Beta for response time inferior or equal to 1s.

235 Some problems that human find relatively easy such as the double keima on  
 236 the second line in problem 16 are difficult for our program, while some prob-  
 237 lems that humans find relatively difficult such as problem 21 are easy for our  
 238 program.

## 239 6. Conclusion

240 We have described an algorithm to detect non transitive connections in the  
 241 game of Go. An optimized Alpha–Beta search is used on top of two General-  
 242 ized Threats Searches. Our program is able to solve problems such as the dou-  
 243 ble monkey jump or the double keima on the second line. It deals with full  
 244 board situations such as the ones encountered in real games.

245 The program can be used with the intersection of the traces for choosing  
246 moves at min nodes. It then solves transitive problems much faster than Al-  
247 pha–Beta and Transitive search with the union of the traces, as can be seen  
248 for problems 2, 5, 8 and 16.

249 It can also be used in the safe mode, taking the union of the traces: even if  
250 the results are not theoretically perfect as can be seen on a pathological posi-  
251 tion, they are pretty reliable given the results on our test suite.

252 The program could be used in a Go program in fast mode, and with the  
253 intersection of the traces, to detect relatively simple non transitivity. The good  
254 point of the transitive search algorithm is that it can solve problems that can-  
255 not be solved by some of the strongest Go programs, the drawback is that it is  
256 much slower than a pattern based approach that only solves the common cases.  
257 The utility of this approach is dependent on the architecture of the program.  
258 Since the algorithm is still slow, it would be difficult to integrate it in a Go pro-  
259 gram based on global search, because the search for transitivity would have to  
260 be done at each call of the global evaluation, unless perhaps the results are ca-  
261 ched. However, it could be used in programs that are not based on global  
262 search, and that spend more time on the evaluation of the position.

263 There is still room for improvements in solving more quickly problems such  
264 as problem 13. It involves improving the connection search algorithm. Future  
265 work also includes extending it toward a more general search program for  
266 combinations of sub-games.

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