

Parity Games: Zielonka’s Algorithm in Quasi-Polynomial Time

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Abstract

Calude, Jain, Khoushainov, Li, and Stephan (2017) proposed a quasi-polynomial-time algorithm solving parity games. After this breakthrough result, a few other quasi-polynomial-time algorithms were introduced; none of them is easy to understand. Moreover, it turns out that in practice they operate very slowly. On the other side there is Zielonka’s recursive algorithm, which is very simple, exponential in the worst case, and the fastest in practice. We combine these two approaches: we propose a small modification of Zielonka’s algorithm, which ensures that the running time is at most quasi-polynomial. In effect, we obtain a simple algorithm that solves parity games in quasi-polynomial time. We also hope that our algorithm, after further optimizations, can lead to an algorithm that shares the good performance of Zielonka’s algorithm on typical inputs, while reducing the worst-case complexity on difficult inputs.

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1 Introduction

The fundamental role of parity games in automata theory and logic, and their applications to verification and synthesis is doubtless, hence it is pointless to elaborate on their importance. Let us only mention that the algorithmic problem of finding the winner in parity games is polynomial-time equivalent to the emptiness problem for nondeterministic automata on infinite trees with parity acceptance conditions, and to the model-checking problem for modal μ -calculus [10]. It also lies at the heart of algorithmic solutions to Church’s synthesis problem [29]. The impact of parity games reaches relatively far areas of computer science, like Markov decision processes [11] and linear programming [15].

It is a long-standing open question whether parity games can be solved in polynomial time. Several results show that they belong to some classes “slightly above” polynomial time. Namely, deciding the winner of parity games was shown to be in $\text{NP} \cap \text{coNP}$ [10], and in $\text{UP} \cap \text{coUP}$ [18], while computing winning strategies is in PLS, PPAD, and even in their subclass CLS [9]. The same holds for other kinds of games: mean-payoff games [36], discounted games, and simple stochastic games [7]; parity games, however, are the easiest among them, in the sense that there are polynomial-time reductions from parity games to the other kinds of games [18, 36], but no reductions in the opposite direction are known.

Describing the algorithmic side of solving parity games, one has to start with Zielonka’s algorithm [35], being an adaptation of an approach proposed by McNaughton to solve Muller



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games [28]. This algorithm consists of a single recursive procedure, being simple and very natural; one may say that it computes who wins the game “directly from the definition”. Its running time is exponential in the worst case [14, 1, 16], but on many typical inputs it works much faster. For over two decades researchers were trying to cutback the complexity of solving parity games, which resulted in a series of algorithms, all of which were either exponential [5, 31, 19, 34, 30, 2], or mildly subexponential [3, 21]. The next era came unexpectedly in 2017 with a breakthrough result of Calude, Jain, Khousainov, Li, and Stephan [6] (see also [17, 23]), who designed an algorithm working in quasi-polynomial time. This invoked a series of quasi-polynomial-time algorithms, which appeared soon after [20, 13, 24]. These algorithms are quite involved (at least compared to the simple recursive algorithm of Zielonka), and it is not so trivial to understand them.

The four quasi-polynomial-time algorithms [6, 20, 13, 24], at first glance being quite different, actually proceed along a similar line (as observed by Bojańczyk and Czerwiński [4] and Czerwiński et al. [8]). Namely, out of all the four algorithms one can extract a construction of a safety automaton (nondeterministic in the case of Lehtinen [24], and deterministic in the other algorithms), which accepts all words encoding plays that are decisively won by one of the players (more precisely: plays consistent with some positional winning strategy), and rejects all words encoding plays in which the player loses (for plays that are won by the player, but not decisively, the automaton can behave arbitrarily). This automaton does not depend at all on the game graph; it depends only on its size. Having an automaton with the above properties, it is not difficult to convert the original parity game into an equivalent safety game (by taking a “product” of the parity game and the automaton), which can be solved easily – and all the four algorithms actually proceed this way, even if it is not stated explicitly that such an automaton is constructed. As shown in Czerwiński et al. [8], all automata having the aforementioned properties have to look very similar: their states have to be leaves of some so-called universal tree; particular papers propose different constructions of these trees, and of the resulting automata (of quasi-polynomial size). Moreover, Czerwiński et al. [8] show a quasi-polynomial lower bound for the size of such an automaton.

In this paper we propose a novel quasi-polynomial-time algorithm solving parity games. It is obtained by applying a small modification to Zielonka’s recursive algorithm; this modification guarantees that the worst-case running time of this algorithm, being originally exponential, becomes quasi-polynomial. The simplicity of Zielonka’s algorithm remains in place; we avoid complicated considerations accompanying all the previous quasi-polynomial-time algorithms. Another point is that our algorithm exploits the structure of parity games in a rather different way from the four previous quasi-polynomial-time algorithms. Indeed, the other algorithms construct automata that are completely independent from a particular game graph given on input – they work in exactly the same way for every game graph of a considered size. The behaviour of our algorithm, in contrast, is highly driven by an analysis of the game graph given on input. In particular, although our algorithm is not faster than quasi-polynomial, it does not fit to the “separator approach” in which a quasi-polynomial lower bound of Czerwiński et al. [8] exists.

The running time of our algorithm is quasi-polynomial, and the space complexity is quadratic (more precisely, $O(n \cdot h)$, where n is the number of nodes in the game graph, and h is the maximal priority appearing there).

We remark that Lehtinen, Schewe, and Wojtczak in their recent follow up paper [25] suggested a variation of our algorithm that improves the complexity to meet the state-of-the-art complexity of broadly $2^{O((\log n)(\log h))}$, while providing polynomial bounds when the number of priorities is logarithmic.

Let us also mention the practical side of the world. It turns out that parity games are one of the areas where theory does not need to meet practice: the quasi-polynomial-time algorithms, although fastest in theory, are actually the slowest. The most exhaustive comparison of existing algorithms was performed by Tom van Dijk [32]. In his Oink tool he has implemented several algorithms, with different optimizations. Then, he has evaluated them on a benchmark of Keiren [22], containing multiple parity games obtained from model checking and equivalence checking tasks, as well as on different classes of random games. It turns out that the classic recursive algorithm of Zielonka [35] performs the best, *ex aequo* with the recent priority promotion algorithm [2]. After that, we have the strategy improvement algorithm [34, 12], being a few times slower. Far later, we have the small progress measure algorithm [19]. At the very end, with a lot of timeouts, we have the quasi-polynomial-time algorithm of Fearnley, Jain, Schewe, Stephan, and Wojtczak [13]. The other quasi-polynomial-time algorithms were not implemented due to excessive memory usage.

While developing the current algorithm, we hoped that it will share the good performance with Zielonka's algorithm, on which it is based. Unfortunately, preliminary experiments have shown that this is not necessarily the case. It turns out that

- on random games our algorithm performs similarly to the slowest algorithms implemented in Oink;
- on crafted game families that are difficult for Zielonka's algorithm, our algorithm is indeed faster from it, but not dramatically faster;
- the only think that is optimistic is that on games with a very low number of priorities our algorithm performs similarly to the fastest algorithms.

Because the empirical results of a direct implementation of the algorithm are completely unsatisfactory, we do not include a full description of our experiments. Instead, we leave an efficient implementation for a future work. Beside of the discouraging outcomes, we believe that our idea, via further optimizations, can lead to an algorithm that is both fast in practice and has a good worst-case complexity (see the concluding section for more comments).

2 Preliminaries

A parity game is played on a *game graph* between two players, called Even or Odd (shortened sometimes to E and O). A game graph consists of

- a directed graph G , where we require that every node has at least one successor, and where there are no self-loops (i.e., edges from a node to itself);
- a labeling of every node v of G by a positive natural number $\pi(v)$, called its *priority*;
- a partition of nodes of G between nodes owned by Even and nodes owned by Odd.

An infinite path in G is called a *play*, while a finite path in G is called a *partial play*. The game starts in a designated starting node. Then, the player to which the current node belongs, selects a successor of this node, and the game continues there. In effect, after a finite time a partial play is obtained, and at the end, after infinite time, this results in a play. We say that a play v_1, v_2, \dots is winning for Even if $\limsup_{i \rightarrow \infty} \pi(v_i)$ is even (i.e., if the maximal priority seen infinitely often is even). Conversely, the play is winning for Odd if $\limsup_{i \rightarrow \infty} \pi(v_i)$ is odd.

A *strategy* of player $P \in \{\text{Even}, \text{Odd}\}$ is a function that maps every partial play that ends in a node of P to some its successor. Such a function says how P will play in every situation of the game (depending on the history of that game). When a (partial) play π follows a strategy σ in every step in which player P is deciding, we say that π *agrees* with σ .

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A strategy σ is *winning* for P from a node v if every play that starts in v and agrees with σ is winning for P . While saying “player P wins from a node v ” we usually mean that P has a winning strategy from v . Let $Win_P(G)$ be the set of nodes of G from which P wins; it is called the *winning region* of P . By Martin's theorem [27] we know that parity games are determined: in every game graph G , and for every node v of G either Even wins from v , or Odd wins from v . In effect, $Win_E(G)$ and $Win_O(G)$ form a partition of the node set of G .

During the analysis, we also consider games with other winning conditions. A *winning condition* is a set of plays. The winning conditions of Even and Odd considered in parity games are denoted $LimsupEven$ and $LimsupOdd$, respectively. Beside of that, for every set S of nodes, let $Safety(S)$ be the set of plays that use only nodes from S .

A *dominion* for Even is a set S of nodes such that from every $v \in S$ Even wins the game with the condition $LimsupEven \cap Safety(S)$; in other words, from every node of S he can win the parity game without leaving S . Likewise, a dominion for Odd is a set S of nodes such that from every $v \in S$ Odd wins the game with the condition $LimsupOdd \cap Safety(S)$. Notice that the whole $Win_P(G)$ is a dominion for P (where $P \in \{Even, Odd\}$). Indeed, if Even is going to win from some $v \in Win_E(G)$, the play cannot leave $Win_E(G)$ and enter a node $v' \in Win_O(G)$, as then Odd could use his winning strategy from v' and win the whole game; here we use the fact that all suffixes of a play in $LimsupEven$ are also in $LimsupEven$. For $P = Odd$ the situation is symmetric.

3 Standard Zielonka's Algorithm

Before presenting our algorithm, we recall the standard Zielonka's algorithm, as a reference.

For a set of nodes N in a game graph G , and for a player $P \in \{Even, Odd\}$, we define the *attractor* of N , denoted $ATR_P(G, N)$, to be the set of nodes of G from which P can force to reach a node from N . In other words, $ATR_P(G, N)$ is the smallest set such that

- $N \subseteq ATR_P(G, N)$,
- if v is a node of P and some its successor is in $ATR_P(G, N)$, then $v \in ATR_P(G, N)$, and
- if v is a node of the opponent of P and all its successors are in $ATR_P(G, N)$, then $v \in ATR_P(G, N)$.

Clearly $ATR_P(G, N)$ can be computed in time proportional to the size of G .

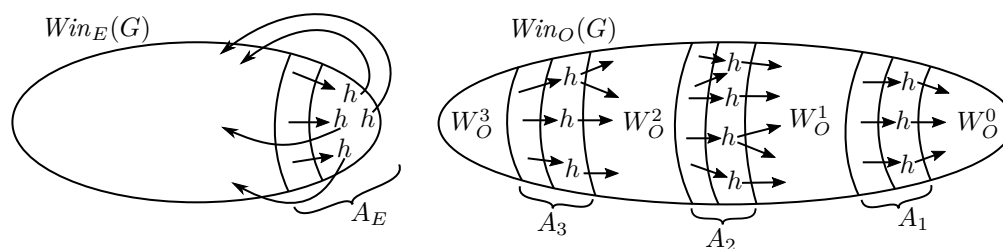
■ **Algorithm 1** Standard Zielonka's Algorithm.

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1: procedure SOLVE $_E(G, h)$                                 ▷  $h$  is an even upper bound for priorities in  $G$ 
2: begin
3:   do begin
4:      $N_h = \{v \in \text{nodes}(G) \mid \pi(v) = h\}$ ;                ▷ nodes with the highest priority
5:      $H = G \setminus ATR_E(G, N_h)$ ;                            ▷ new game: reaching priority  $h \rightarrow$  win
6:      $W_O = \text{SOLVE}_O(H, h - 1)$ ;                            ▷ in  $W_O$  we lose before reaching priority  $h$ 
7:      $G = G \setminus ATR_O(G, W_O)$ ;                            ▷ possibly  $N_h \cap ATR_O(G, W_O) \neq \emptyset$ 
8:   end while  $W_O \neq \emptyset$ ;
9:   return  $\text{nodes}(G)$ ;
10: end

```

Algorithm 1 is the standard Zielonka's algorithm. The procedure $\text{SOLVE}_E(G, h)$ returns $Win_E(G)$, the winning region of Even, if h is an even number that is greater or equal than all priorities appearing in G . A procedure $\text{SOLVE}_O(G, h)$ is also needed; it is identical to $\text{SOLVE}_E(G, h)$ except that the roles of E and O are swapped; it returns $Win_O(G)$, the



■ **Figure 1** The structure of winning regions in a parity game.

winning region of Odd. While writing $G \setminus S$, we mean the game obtained by removing from G all nodes in S , and all edges leading to nodes in S or starting from nodes in S . We use this construct only when S is an attractor; in such a case, if all successors of a node v are removed, then v is also removed (i.e., if all successors of v belong to an attractor, then v belongs to the attractor as well). In effect $G \setminus S$ is a valid game graph (every its node has at least one successor).

We remark that the algorithm is presented in a slightly different way than usually. Namely, we use here a loop, while the usual presentation does not use a loop but rather calls recursively $\text{SOLVE}_E(G \setminus \text{ATR}_O(G, W_O), h)$ at the end of the procedure. This is only a superficial difference in the presentation, but is useful while modifying the algorithm in the next section.

The algorithm can be understood while looking at Figure 1. Let h be the highest priority used in G ; assume that it is even. The game graph G can be divided into two parts: $\text{Win}_E(G)$ and $\text{Win}_O(G)$. In $\text{Win}_E(G)$ we can distinguish the attractor of nodes with priority h (denoted A_E). Odd either loses inside $\text{Win}_E(G) \setminus A_E$, or enters A_E , which causes that a node with priority h is seen, and then the game continues in some node of $\text{Win}_E(G)$. The winning region of Odd, $\text{Win}_O(G)$, can be divided into multiple parts. We have a part W_O^0 , where Odd can win without seeing a node of priority h . Then, we have nodes of priority h from which Even is forced to enter W_O^0 , and their attractor, denoted A_1 . Then, we have a part W_O^1 , where Odd can ensure that the play is either winning for him inside W_O^1 or enters A_1 ; in other words, from nodes of W_O^1 Odd can win while seeing h at most once. We also have parts W_O^i for larger i , and corresponding attractors A_i .

While running the algorithm, this partition of G is not known, and has to be discovered. To this end, the algorithm assumes first (in the game H) that all nodes of priority h are winning for Even. The first call to $\text{SOLVE}_O(H, h - 1)$ returns the set W_O^0 of nodes where Odd wins without seeing a node of priority h . We then remove them from the game, together with the attractor A_1 . In the next step, $\text{SOLVE}_O(H, h - 1)$ returns the set W_O^1 , and so on. At the end the whole $\text{Win}_O(G)$ becomes removed, and the procedure returns $\text{Win}_E(G)$.

4 Quasi-Polynomial-Time Algorithm

We now present a modification to Algorithm 1 that results in obtaining quasi-polynomial running time, in the worst case.

The modification can be understood while looking again at Figure 1. The key observation is that, while $\text{Win}_O(G)$ is of size at most n (where n is the number of nodes in G), then most of its parts W_O^i are smaller. Namely, most of them have to be of size at most $\frac{n}{2}$, and only one of them can be larger than $\frac{n}{2}$. We use this observation, and while looking for W_O^i , we search for a winning region (for a dominion) of size at most $\frac{n}{2}$. Usually this is enough;

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only once it is not enough: one W_O^i can be larger than $\frac{n}{2}$ and it will not be found if we only look for a set of size at most $\frac{n}{2}$. But when the algorithm finds no set of size at most $\frac{n}{2}$, we can once search for W_O^i of an arbitrary size. After that, we know that all following sets W_O^i are again of size at most $\frac{n}{2}$. While going recursively, we notice that every W_O^i can be further subdivided in a similar way, while splitting on the priority $h - 2$. If $|W_O^i| \leq \frac{n}{2}$, we again have the property that most of the parts of W_O^i are of size at most $\frac{n}{4}$, and only one of them can be larger than $\frac{n}{4}$.

To exploit this observation, in the recursive calls we pass two precision parameters, p_E and p_O (one for every of the players), saying that we search for winning sets of size at most p_E for Even, and at most p_O for Odd. The modified procedure is presented as Algorithm 2. Again, one also needs a procedure SOLVE_O , which is obtained from SOLVE_E by literally changing every E to O and *vice versa*.

■ Algorithm 2 Quasi-Polynomial-Time Algorithm.

```

1: procedure  $\text{SOLVE}_E(G, h, p_E, p_O)$  ▷  $p_E, p_O$  are new “precision” parameters
2: begin
3:   if  $G = \emptyset \vee p_E \leq 1$  then
4:     return  $\emptyset$ ; ▷ we assume that there are no self-loops in  $G$ 
5:   do begin
6:      $N_h = \{v \in \text{nodes}(G) \mid \pi(v) = h\}$ ;
7:      $H = G \setminus \text{ATR}_E(G, N_h)$ ;
8:      $W_O = \text{SOLVE}_O(H, h - 1, \lfloor p_O/2 \rfloor, p_E)$ ; ▷ precision decreased
9:      $G = G \setminus \text{ATR}_O(G, W_O)$ ;
10:  end while  $W_O \neq \emptyset$ ;
11:   $N_h = \{v \in \text{nodes}(G) \mid \pi(v) = h\}$ ;
12:   $H = G \setminus \text{ATR}_E(G, N_h)$ ;
13:   $W_O = \text{SOLVE}_O(H, h - 1, p_O, p_E)$ ; ▷ we try once with the full precision
14:   $G = G \setminus \text{ATR}_O(G, W_O)$ ;
15:  while  $W_O \neq \emptyset$  do begin
16:     $N_h = \{v \in \text{nodes}(G) \mid \pi(v) = h\}$ ;
17:     $H = G \setminus \text{ATR}_E(G, N_h)$ ;
18:     $W_O = \text{SOLVE}_O(H, h - 1, \lfloor p_O/2 \rfloor, p_E)$ ; ▷ again, precision decreased
19:     $G = G \setminus \text{ATR}_O(G, W_O)$ ;
20:  end;
21:  return  $\text{nodes}(G)$ ;
22: end

```

We start the algorithm with $p_E = p_O = n$, where n is the number of nodes in G . In the procedure we have now, in a sense, three copies of the previous procedure, corresponding to three stages. In the first stage, in lines 5-10, we look for sets W_O^i of size at most $\lfloor \frac{p_O}{2} \rfloor$. If the returned set is empty, this may mean that the next W_O^i either is empty, or is of size greater than $\lfloor \frac{p_O}{2} \rfloor$. Then, in lines 11-14, we once search for a set W_O^i of size at most p_O (knowing that if it is nonempty, then its size is greater than $\lfloor \frac{p_O}{2} \rfloor$). Finally, in the loop in lines 15-20, we again look for sets W_O^i of size at most $\lfloor \frac{p_O}{2} \rfloor$ (because we have already found a set of size greater than $\lfloor \frac{p_O}{2} \rfloor$, all the remaining sets have size at most $\lfloor \frac{p_O}{2} \rfloor$).

5 Complexity Analysis

Let us analyze the complexity of our algorithm.

First, we observe that the space complexity is $O(n \cdot h)$, where n is the number of nodes, and h is the maximal priority. Indeed, the depth of the recursion is at most h , and on every step we only need to remember some sets of nodes.

We now come to the running time. As it is anyway worse than the running time of the other quasi-polynomial-time algorithms, we do not aim in proving a very tight upper bound; we only prove that the running time is quasi-polynomial.

Let $R(h, l)$ be the number of (nontrivial) executions of the SOLVE_E and SOLVE_O procedures performed during one call to $\text{SOLVE}_E(G, h, p_E, p_O)$ with $\lfloor \log p_E \rfloor + \lfloor \log p_O \rfloor = l$, and with G having at most n nodes (where n is fixed). We only count here nontrivial executions, that is, such that do not leave the procedure in line 4. Clearly $R(0, l) = R(h, 0) = 0$. For $h, l \geq 1$ it holds that

$$R(h, l) \leq 1 + n \cdot R(h - 1, l - 1) + R(h - 1, l). \quad (1)$$

Indeed, in SOLVE_E after every call to SOLVE_O we remove at least one node from G , with the exception of two such calls: the last call in line 8, and the last call ever. In effect, in lines 8 and 18 we have at most n calls to SOLVE_O with decreased precision (plus, potentially, the $(n + 1)$ -th call with empty G , which is not included in $R(h, l)$), and in line 13 we have one call to SOLVE_O with full precision. Notice that $\lfloor \log p_O \rfloor$ (hence also l) decreases by 1 in the decreased-precision call.

Using Inequality (1) we now prove by induction that $R(h, l) \leq n^l \cdot \binom{h+l}{l} - 1$. For $h = 0$ and for $l = 0$ the inequality holds. For $h, l \geq 1$ we have that

$$\begin{aligned} R(h, l) &\leq 1 + n \cdot R(h - 1, l - 1) + R(h - 1, l) \\ &\leq 1 + n \cdot \left(n^{l-1} \cdot \binom{h-1+l-1}{l-1} - 1 \right) + n^l \cdot \binom{h-1+l}{l} - 1 \\ &\leq n^l \cdot \left(\binom{h-1+l}{l-1} + \binom{h-1+l}{l} \right) - 1 \\ &= n^l \cdot \binom{h+l}{l} - 1. \end{aligned}$$

In effect, $R(h, l) \leq n^l \cdot (h + l)^l$. Recalling that we start with $l = 2 \cdot \lfloor \log n \rfloor$, we see that this number is quasi-polynomial in n and h . This concludes the proof, since obviously a single execution of the SOLVE_E procedure (not counting the running time of recursive calls) costs polynomial time.

6 Correctness

We now justify correctness of the algorithm. This amounts to proving the following lemma.

► **Lemma 6.1.** *Procedure $\text{SOLVE}_E(G, h, p_E, p_O)$ returns a set W_E such that for every $S \subseteq \text{nodes}(G)$,*

- *if S is a dominion for Even, and $|S| \leq p_E$, then $S \subseteq W_E$, and*
- *if S is a dominion for Odd, and $|S| \leq p_O$, then $S \cap W_E = \emptyset$.*

Notice that in G there may be nodes that do not belong to any dominion smaller than p_E or p_O ; for such nodes we do not specify whether or not they are contained in W_E .

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Recall that $Win_E(G)$ is a dominion for Even, and $Win_O(G)$ is a dominion for Odd. Thus, using Lemma 6.1 we can conclude that for $p_E = p_O = n$ the procedure returns $Win_E(G)$, the winning region of Even.

One may wonder why we use dominions in the statement of the lemma, instead of simply saying that if $|Win_E(G)| \leq p_E$, then $Win_E(G) \subseteq W_E$. Such a simplified statement, however, is not suitable for induction. Indeed, while switching from the game G to the game H (created in lines 7, 12, 17) the winning regions of Even may increase dramatically, because in H Odd is not allowed to visit any node with priority h . Nevertheless, the winning region of Even in G , and any dominion of Even in G , remains a dominion in H (when restricted to nodes of H).

Before proving Lemma 6.1, let us observe two facts about dominions. In their statements $P \in \{\text{Even}, \text{Odd}\}$ is one of the players, and \bar{P} is his opponent.

► **Fact 6.2.** *If S is a dominion for P in a game G , and X is a set of nodes of G , then $S \setminus \text{ATR}_P(G, X)$ is a dominion for P in $G \setminus \text{ATR}_P(G, X)$.*

Proof. Denote $S' = S \setminus \text{ATR}_P(G, X)$ and $G' = G \setminus \text{ATR}_P(G, X)$. By definition, from every node $v \in S$ player P wins with the condition $\text{Limsup}P \cap \text{Safety}(S)$ in G , using some winning strategy. Observe that using the same strategy he wins with the condition $\text{Limsup}P \cap \text{Safety}(S')$ in G' (assuming that the starting node v is in S'). The strategy remains valid in G' , because every node u of player P that remains in G' has the same successors in G' as in G (conversely: if some of successors of u belongs to $\text{ATR}_P(G, X)$, then u also belongs to $\text{ATR}_P(G, X)$). ◀

► **Fact 6.3.** *If S is a dominion for P in a game G , and X is a set of nodes of G such that $S \cap X = \emptyset$, then S is a dominion for P in $G \setminus \text{ATR}_{\bar{P}}(G, X)$ (in particular $S \subseteq \text{nodes}(G \setminus \text{ATR}_{\bar{P}}(G, X))$).*

Proof. Denote $G' = G \setminus \text{ATR}_{\bar{P}}(G, X)$. Suppose that there is some $v \in S \cap \text{ATR}_{\bar{P}}(G, X)$. On the one hand, P can guarantee that, while starting from v , the play stays in S (by the definition of a dominion); on the other hand, \bar{P} can force to reach the set X (by the definition of an attractor), which is disjoint from S . Thus such a node v could not exist, we have $S \subseteq \text{nodes}(G')$.

It remains to observe that from every node $v \in S$ player P wins with the condition $\text{Limsup}P \cap \text{Safety}(S)$ also in the restricted game G' , using the same strategy as in G . Indeed, a play in G following this strategy never leaves S , and the whole S remains unchanged in G' . ◀

We are now ready to prove Lemma 6.1.

Proof of Lemma 6.1. We prove the lemma by induction on h . Consider some execution of the procedure. By G^i, N_h^i, H^i, W_O^i we denote values of the variables G, N_h, H, W_O just after the i -th call to SOLVE_O in one of the lines 8, 13, 18; in lines 9, 14, 19 we create G^{i+1} out of G^i and W_O^i . In particular G^1 equals the original game G , and at the end we return $\text{nodes}(G^{m+1})$, where m is the number of calls to SOLVE_O .

Concentrate on the first item of the lemma: fix an Even's dominion S in G (i.e., in G^1) such that $|S| \leq p_E$. Assume that $S \neq \emptyset$ (for $S = \emptyset$ there is nothing to prove). Notice first that a nonempty dominion has at least two nodes (by assumption there are no self-loops in G , hence every play has to visit at least two nodes), thus, because $S \subseteq \text{nodes}(G)$ and $|S| \leq p_E$, we have that $G \neq \emptyset$ and $p_E > 1$. It means that the procedure does not return in line 4. We thus need to prove that $S \subseteq \text{nodes}(G^{m+1})$.

We actually prove that S is a dominion for Even in G^i for every $i \in \{1, \dots, m+1\}$, meaning in particular that $S \subseteq \text{nodes}(G^i)$. This is shown by an internal induction on i . The base case ($i = 1$) holds by assumption. For the induction step, consider some $i \in \{1, \dots, m\}$. By the induction assumption S is a dominion for Even in G^i , and we need to prove that it is a dominion for Even in G^{i+1} .

Consider $S^i = S \cap \text{nodes}(H^i)$. Because $S^i = S \setminus \text{ATR}_E(G^i, N_h^i)$, by Fact 6.2 the set S^i is a dominion for Even in $H^i = G^i \setminus \text{ATR}_E(G^i, N_h^i)$, and obviously $|S^i| \leq |S| \leq p_E$. By the assumption of the external induction (which can be applied to SOLVE_O , by symmetry) it follows that $S^i \cap W_O^i = \emptyset$, so also $S \cap W_O^i = \emptyset$ (because W_O^i contains only nodes of G^i , while $S \setminus S^i$ contains no nodes of G^i). Thus, by Fact 6.3 the set S is a dominion for Even in $G^{i+1} = G^i \setminus \text{ATR}_O(G^i, W_O^i)$. This finishes the proof of the first item.

Now we prove the second item of the lemma. To this end, fix some Odd's dominion S in G such that $|S| \leq p_O$. If $p_E \leq 1$, we return $W_E = \emptyset$ (line 4), so clearly $S \cap W_E = \emptyset$. The interesting case is when $p_E \geq 2$. Denote $S^i = S \cap \text{nodes}(G^i)$ for all $i \in \{1, \dots, m+1\}$; we first prove that S^i is a dominion for Odd in G^i . This is shown by induction on i . The base case of $i = 1$ holds by assumption, because $G^1 = G$ and $S^1 = S$. For the induction step, assume that S^i is a dominion for Odd in G^i , for some $i \in \{1, \dots, m\}$. By definition $G^{i+1} = G^i \setminus \text{ATR}_O(G^i, W_O^i)$ and $S^{i+1} = S^i \setminus \text{ATR}_O(G^i, W_O^i)$, so S^{i+1} is a dominion for Odd in G^{i+1} by Fact 6.2, which finishes the inductive proof.

For $i \in \{1, \dots, m\}$, let Z^i be the set of nodes (in $S^i \setminus N_h^i$) from which Odd wins with the condition $\text{LimsupOdd} \cap \text{Safety}(S^i \setminus N_h^i)$ in G^i (that is, where Odd can win without seeing priority h – the highest even priority). Let us observe that if $S^i \neq \emptyset$ then $Z^i \neq \emptyset$ (\clubsuit). Indeed, suppose to the contrary that $Z^i = \emptyset$, and consider an Odd's strategy that allows him to win with the condition $\text{LimsupOdd} \cap \text{Safety}(S^i)$ in G^i , from some node $v_0 \in S^i$. Because $v_0 \notin Z^i$, this strategy is not winning for the condition $\text{LimsupOdd} \cap \text{Safety}(S^i \setminus N_h^i)$, so Even, while playing against this strategy, can reach a node v_1 in N_h^i (as he cannot violate the parity condition nor leave S^i). For the same reason, because $v_1 \notin Z^i$, Even can continue and reach a node v_2 in N_h^i . Repeating this forever, Even gets priority h (which is even and is the highest priority) infinitely many times, contradicting the fact that the strategy was winning for Odd.

Observe also that from nodes of Z^i Odd can actually win with the condition $\text{LimsupOdd} \cap \text{Safety}(Z^i)$ in G^i , using the strategy that allows him to win with the condition $\text{LimsupOdd} \cap \text{Safety}(S^i \setminus N_h^i)$. Indeed, if a play following this strategy enters some node v , then from this node v Odd can still win with the condition $\text{LimsupOdd} \cap \text{Safety}(S^i \setminus N_h^i)$, which means that these nodes belongs to Z^i . It follows that Z^i is a dominion for Odd in G^i . Moreover, because $Z^i \cap N_h^i = \emptyset$, from Fact 6.3 we have that Z^i is a dominion for Odd in $H^i = G^i \setminus \text{ATR}_E(G^i, N_h^i)$.

Let k be the number of the call to SOLVE_O that is performed in line 13 (calls number $1, \dots, k-1$ are performed in line 8, and calls number $k+1, \dots, m$ are performed in line 18). Recall that W_O^i is the set returned by a call to $\text{SOLVE}_O(H^i, h-1, p_O^i, p_E)$, where $p_O^k = p_O$, and $p_O^i = \lfloor \frac{p_O}{2} \rfloor$ if $i \neq k$. From the assumption of the external induction, if $|Z^i| \leq \lfloor \frac{p_O}{2} \rfloor$ or if $i = k$ (since $Z^i \subseteq S^i \subseteq S$ and $|S| \leq p_O$, clearly $|Z^i| \leq p_O$), we obtain that $Z^i \subseteq W_O^i$ (\spadesuit).

We now prove that $|S^{k+1}| \leq \lfloor \frac{p_O}{2} \rfloor$. This clearly holds if $S^{k-1} = \emptyset$, because $S^{k+1} \subseteq S^k \subseteq S^{k-1}$. Suppose thus that $S^{k-1} \neq \emptyset$. Then $Z^{k-1} \neq \emptyset$, by (\clubsuit). On the other hand, $W_O^{k-1} = \emptyset$, because we are just about to leave the loop in lines 5-10 (the k -th call to SOLVE_O is in line 13). By (\spadesuit), if $|Z^{k-1}| \leq \lfloor \frac{p_O}{2} \rfloor$, then $Z^{k-1} \subseteq W_O^{k-1}$, which does not hold in our case. Thus $|Z^{k-1}| > \lfloor \frac{p_O}{2} \rfloor$. Because $W_O^{k-1} = \emptyset$, we simply have $G^k = G^{k-1}$, and $S^k = S^{k-1}$, and $Z^k = Z^{k-1}$. Using (\spadesuit) for $i = k$, we obtain that $Z^k \subseteq W_O^k$, and because $S^{k+1} = S^k \setminus \text{ATR}_O(G^k, W_O^k) \subseteq S^k \setminus W_O^k \subseteq S^k \setminus Z^k$ we obtain that $|S^{k+1}| \leq |S^k| - |Z^k| \leq p_O - (\lfloor \frac{p_O}{2} \rfloor + 1) \leq \lfloor \frac{p_O}{2} \rfloor$, as initially claimed.

If $k = m$, we have $Z^m \subseteq W_O^m$ by (\spadesuit). If $k + 1 \leq m$, we have $S^m \subseteq S^{k+1}$ (our procedure only removes nodes from the game) and $Z^m \subseteq S^m$, so $|Z^m| \leq \lfloor \frac{p_O}{2} \rfloor$ by the above paragraph, and also $Z^m \subseteq W_O^m$ by (\spadesuit). Because after the m -th call to SOLVE_O the procedure ends, we have $W_O^m = \emptyset$, so also $Z^m = \emptyset$, and thus $S^m = \emptyset$ by (\clubsuit). We have $S^{m+1} \subseteq S^m$, so $S^{m+1} = S \cap \text{nodes}(G^{m+1}) = \emptyset$. This is exactly the conclusion of the lemma, since the set returned by the procedure is $\text{nodes}(G^{m+1})$. \blacktriangleleft

7 Conclusions

To the list of the four existing quasi-polynomial-time algorithms solving parity games, we have added a new one. It uses a rather different approach: it analyses recursively the game graph, like Zielonka's algorithm.

Notice that the number of recursive calls in our algorithm may be smaller than in the original Zielonka's algorithm, because of the precision parameters, but it may also be larger. Indeed, while SOLVE_E in the original Zielonka's algorithm stops after the first time when a recursive call returns \emptyset , in our algorithm the procedure stops after the second time when a recursive call returns \emptyset .

The algorithm, as is, turns out not to be very efficient in practice. Beside of that, we believe that it can serve as a good starting point for a more optimized algorithm. Over the years, some optimizations to Zielonka's algorithm were proposed. For example, Liu, Duan, and Tian [26] replace the loop guard $W_O = \emptyset$ by $W_O = \text{ATR}_O(G, W_O)$ (which ensures that W_O will be empty in the next iteration of the loop). Verver [33] proposes to check whether $\text{ATR}_E(G, N_h)$ contains all nodes of priority $h - 1$, and if so, to extend N_h by nodes of the next highest Even priority (i.e., $h - 2$). It seems that these optimizations can be applied to our algorithm as well.

A straightforward optimization is to decrease p_O and p_E to $|G|$ at the beginning of every recursive call.

Another idea is to extend the recursive procedure so that it will return also a Boolean value saying whether the returned set surely equals the whole winning region (i.e., whether the precision parameters have not restricted anything). If while making the recursive call with smaller precision (line 8) the answer is positive, but the returned set W_O is empty, we can immediately stop the procedure, without making the recursive call with the full precision (line 13).

One can also observe that the call to SOLVE_O in line 13 (with the full precision) gets the same subgame H as the last call to SOLVE_O in line 8 (with decreased precision). A very rough idea is to make some use of the computations performed by the decreased-precision call during the full-precision call.

We leave implementation and evaluation of the above (and potentially some other) optimizations for a future work.

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