

The priority promotion approach to parity games

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Abstract. We consider *parity games*, a special form of two-player infinite-duration games on numerically labeled graphs, whose winning condition requires that the maximal value of a label occurring infinitely often during a play be of some specific parity. We present a new family of algorithms for the solution of the problem, based on the idea of promoting vertices to higher priorities during the search for winning regions. The proposed approach has nice computational properties, exhibiting the best space complexity among the currently known solutions. Experimental results on both random games and benchmark families show that the technique is also very effective in practice.

1 Introduction

Parity games are perfect-information two-player turn-based games of infinite duration, usually played on finite directed graphs. Their vertices, called *positions*, are labeled by natural numbers, called *priorities*, and are assigned to one of two players, named *Even* and *Odd* or, simply, 0 and 1, respectively. The game starts at an arbitrary position and, during its evolution, players can take a move (an outgoing edge) only at their own positions. The moves selected by the players induce an infinite sequence of vertices, called *play*. If the maximal priority of the vertices occurring infinitely often in the play is *even*, then the play is *winning* for player 0, otherwise, player 1 takes it all. The importance of these games is due to the numerous applications in automata theory and in the area of system specification, verification, and synthesis, where it is used as algorithmic back-end of satisfiability and model-checking procedures for temporal logics automata theory. In particular, it has been proved to be linear-time interreducible with the model-checking problem for the *modal* μ CALCULUS [6]. Besides the practical importance, parity games are also interesting from a computational complexity point of view, since their solution problem is one of the few inhabitants of the $\text{UPTIME} \cap \text{COUPTIME}$ class [8] for which it is still open is the question about the membership in PTIME . The literature on the topic is rich of algorithms for solving parity games, which can be mainly classified into two families. The first one contains the algorithms that recursively decompose the problem into subproblems, whose solutions are then suitably assembled to obtain the desired result. In this category fall, for example, *Zielonka’s recursive algorithm* [14] and

its *dominion decomposition* [11] and *big step* [12] improvements. The second family, instead, groups together those algorithms that compute a winning strategy for the two players on the entire game, *e.g.*, the *Jurdziński’s progress measure* algorithm [9] and the *strategy improvement* approach [13]. Recently, [5] has shown that the problem can be solved in quasi-polynomial time. Despite the theoretical relevance of the result, it does not seem obvious how to turn the employed technique into practical algorithms. The first two implementations based on this result (see [10] and [7]) are no match for any of the exponential algorithms.

This extended abstract presents the ideas underlying a new family of algorithms for solving parity games described in [2–4], based on the notions of *quasi dominion* and *priority promotion*. A quasi dominion Q for player $\alpha \in \{0, 1\}$, called a *quasi α -dominion*, is a set of vertices from each of which player α can enforce a winning play that never leaves the region, unless one of the following two conditions holds: (i) the opponent $\bar{\alpha}$ can escape from Q or (ii) the only choice for player α itself is to exit from Q . Quasi dominions of the same player can be merged together to form larger quasi α -dominions until, eventually, a dominion for some player is found. The resulting technique, while still exhibiting exponential behaviors, only requires $O(n \cdot \log k)$ memory (with n and k the number of positions and priorities), which improves on the state-of-the-art. Experimental results show that the proposed approach performs very well in practice, most often significantly better than existing ones, as extensively shown in [1–4].

2 The priority promotion approach

For the sake of space, we refer to [4] for the formal definitions of parity game, dominion, (winning) strategy, and attractor.

Priority promotion. The priority promotion approach attacks the problem of solving a parity game \mathcal{D} by computing one of its dominions D , for some player α , at a time. Indeed, once the α -attractor D^* of D is removed from \mathcal{D} , the smaller game $\mathcal{D} \setminus D^*$ is obtained, whose positions are winning for one player iff they are winning for the same player in the original game. This allows for decomposing the problem of solving a parity game into iteratively finding its dominions [11]. In order to solve the dominion problem, the idea is to start from a much weaker notion than that of dominion, called *quasi dominion*. Intuitively, a quasi α -dominion is a set of positions on which player α has a strategy whose induced plays either remain inside the set forever and are winning for α or can exit from it passing through the set of escape positions.

Definition 1 (Quasi Dominion [4]). *A non-empty set of positions Q is a quasi α -dominion in \mathcal{D} for a player α if there exists an α -strategy whose induced plays are either infinite and winning for α or finite and end in an $\bar{\alpha}$ -escaping position.*

If all induced plays remain in the set Q forever, then we have an α -dominion, *i.e.*, a subset of the α -winning region. In this case, the escape set of Q is empty and we say that Q is *α -closed*. Otherwise, we say that it is *α -open*.

The priority promotion algorithm explores a partial order $\langle S, \prec \rangle$, whose elements, called *states*, record information about the open quasi dominions computed along the way. In the initial state $\top \in S$ the quasi dominions are initialized to the sets of positions with the same priority. At each step, a new quasi dominion is extracted from the current state, by means of a *query* operator \mathfrak{R} , and used to compute a successor state, by means of a *successor* operator \downarrow , if the quasi dominion is open. If, on the other hand, it is closed, the search is over. The partial order $\langle S, \top, \prec \rangle$ together with the operators \mathfrak{R} and \downarrow form what is called a *dominion space*. The notion of dominion space is quite general and can be instantiated in different ways, by providing specific query and successor operators (see [2–4]). The overall procedure is sound and complete on any dominion space and its time complexity is linear in the *execution depth* of the space, namely the length of the longest chain in the underlying partial order compatible with the successor operator. Its space complexity, instead, is only logarithmic in the *size* of the dominion-space, since only one state at the time needs to be maintained.

The priority promotion algorithm processes the input game in descending order of priority and, at each step, a subgame of the entire game, obtained by removing the quasi dominions previously computed at higher priorities, is considered. At each priority of parity α : (i) a quasi α -domain Q is extracted by the query operator from the current subgame; (ii) if Q is closed in the entire game, the search stops and returns Q as result; otherwise, (iii) a successor state in the underlying partial order is computed by the successor operator, depending on whether Q is open in the current subgame or not. In the first case, the quasi α -dominion is removed from the current subgame and the search restarts on the new subgame that can only contain positions with lower priorities. In the second case, Q is merged together with some previously computed quasi α -dominion with higher priority. Being a dominion space well-ordered, the search is guaranteed to eventually terminate and return a closed quasi dominion. The procedure requires the solution of two crucial problems: (a) *extracting a quasi dominion* from a subgame and (b) *merging together two quasi α -dominions*.

The solution of the first problem relies on the definition of a specific class of quasi dominions, called *regions*. An α -region R of a game \mathcal{D} is a quasi α -dominion of \mathcal{D} all of whose escape positions have the maximal priority $p \triangleq \text{pr}(\mathcal{D}) \equiv_2 \alpha$ in \mathcal{D} . In this case, we say that the α -region R has priority p . If player $\bar{\alpha}$ escapes from α -region R , it must visit a position with the highest priority in it and parity α .

Definition 2 (Region [4]). *A quasi α -dominion R is an α -region in \mathcal{D} if $\text{pr}(\mathcal{D}) \equiv_2 \alpha$ and all its $\bar{\alpha}$ -escape positions have priority $\text{pr}(\mathcal{D})$.*

It is important to observe that, in any parity game, an α -region always exists, for some $\alpha \in \{0, 1\}$. In particular, the set of positions of maximal priority in the game always forms an α -region, with α equal to the parity of that maximal priority. A closed α -region in a game is clearly an α -dominion in that game. In addition, the α -attractor of an α -region is always an α -region. These observations give us an easy and efficient way to extract a quasi dominion from every subgame: collect the α -attractor of the positions with maximal priority p in the subgame,

where $p \equiv_2 \alpha$, and assign p as priority of the resulting region R . This priority, called *measure* of R , intuitively corresponds to an under-approximation of the best priority player α can force the opponent $\bar{\alpha}$ to visit along any play exiting from R . During the search for a dominion, the computed regions, together with their current measure, are kept track of by means of an auxiliary priority function $r : \text{Ps} \rightarrow \text{Pr}$, called *region function*. Given a priority p , we denote by $r^{(\geq p)}$ (*resp.*, $r^{(>p)}$ and $r^{(<p)}$) the function obtained by restricting the domain of r to the positions with measure greater than or equal to p (*resp.*, greater than and lower than p). Moreover, the set of pairs (R, α) , where R is an α -region, is denoted by Rg , and is partitioned into the sets Rg^- and Rg^+ of open and closed α -region pairs, respectively.

Algorithm 1 provides the implementation for the query function compatible with the priority-promotion mechanism. Line 1 simply computes the parity α of the priority to process in the state $s \triangleq (r, p)$. Line 2, instead, computes the attractor *w.r.t.* player α in subgame $\mathcal{D}_s \triangleq \mathcal{D} \setminus \text{dom}(r^{(>p)})$ of the region contained in r at the current priority p . The resulting set R is an α -region of \mathcal{D}_s containing $r^{-1}(p)$ (see [4]).

A solution to the second problem, the merging operation, is obtained as follows. Given an α -region R in some game \mathcal{D} and an α -dominion D in a subgame of \mathcal{D} that does not contain R itself, the two sets are merged together, if the only moves exiting from $\bar{\alpha}$ -positions of D in the entire game lead to higher priority α -regions and R has the lowest priority among them. The priority of R is called the *best escape priority* of D for $\bar{\alpha}$. The correctness of this merging operation is established in [4].

The merging operation is implemented by promoting all the positions of α -dominion D to the measure of R , thus improving the measure of D . For this reason, it is called a *priority promotion*. After a promotion to some measure p , the regions with measure lower than p might need to be destroyed, by resetting all the contained positions to their original priority (see [4]). The reset is in general unavoidable, since the new promoted region may attract positions from lower ones, thereby potentially invalidating their status as regions. In some cases the player who wins by remaining in the region may even change from α to $\bar{\alpha}$. The promotion operation is based on the notion of best escape priority mentioned above, namely the priority of the lowest α -region in r that has an

Algorithm 1: Query Function.

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signature  $\mathfrak{R}_{\mathcal{D}} : S_{\mathcal{D}} \rightarrow 2^{\text{Ps}_{\mathcal{D}}} \times \{0, 1\}$ 
function  $\mathfrak{R}_{\mathcal{D}}(s)$ 
  let  $(r, p) = s$  in
1    $\alpha \leftarrow p \bmod 2$ 
2    $R \leftarrow \text{atr}_{\mathcal{D}_s}^{\alpha}(r^{-1}(p))$ 
3   return  $(R, \alpha)$ 

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Algorithm 2: Successor Function.

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signature  $\downarrow_{\mathcal{D}} : \succ_{\mathcal{D}} \rightarrow \Delta_{\mathcal{D}} \times \text{Pr}_{\mathcal{D}}$ 
function  $s \downarrow_{\mathcal{D}}(R, \alpha)$ 
  let  $(r, p) = s$  in
1   if  $(R, \alpha) \in \text{Rg}_{\mathcal{D}_s}^-$  then
2      $r^* \leftarrow r[R \mapsto p]$ 
3      $p^* \leftarrow \max(\text{rng}(r^{(<p)}))$ 
   else
4      $p^* \leftarrow \text{bep}_{\mathcal{D}}^{\bar{\alpha}}(R, r)$ 
5      $r^* \leftarrow \text{pr}_{\mathcal{D}} \uplus r^{(\geq p^*)}[R \mapsto p^*]$ 
6   return  $(r^*, p^*)$ 

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incoming move from the α -region R , closed in the current subgame, that needs to be promoted. Algorithm 2 reports the pseudo-code of the successor function. Given the current state s and a region pair (R, α) open in the whole game, it produces a successor state $s^* \triangleq (r^*, p^*)$. It first checks whether R is open also in the subgame \mathcal{D}_s (Line 1). If this is the case, it assigns measure p to region R and stores it in the new region function r^* (Line 2) and the next priority p^* is set to the highest priority lower than p (Line 3). Otherwise, a promotion, merging R with some other α -region contained in r , is required. The next priority p^* is set to the best escape priority of R for player $\bar{\alpha}$ in the entire game \mathcal{D} *w.r.t.* r (Line 4). Region R is, then, promoted to priority p^* and all the regions with lower measure than p^* in the region function r are reset by means of the operator $f \uplus g$, which complete the missing values for the domain of g with the values from f .

The result is a sound and complete solution procedure, as stated below.

Theorem 1. *The Priority Promotion algorithm is correct and solves a game with $n \in \mathbb{N}$ positions and $k \in [1, n]$ priorities in $O(n \cdot \log k)$ space.*

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