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# One-Clock Priced Timed Games with Arbitrary Weights ${ }^{\text {T}}$ 

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#### Abstract

Priced timed games are two-player zero-sum games played on priced timed automata (whose locations and transitions are labeled by weights modeling the price of spending time in a state and executing an action, respectively). The goals of the players are to minimise and maximise the price to reach a target location, respectively. We consider priced timed games with one clock and arbitrary integer weights and show that, for an important subclass of theirs (the so-called simple priced timed games), one can compute, in exponential time, the optimal values that the players can achieve, with their associated optimal strategies. As side results, we also show that one-clock priced timed games are determined and that we can use our result on simple priced timed games to solve the more general class of so-called negative-reset-acyclic priced timed games (with arbitrary integer weights and one clock). The decidability status of the full class of priced timed games with one-clock and arbitrary integer weights still remains open.


Keywords: Priced timed games, Real-time systems, Game theory

## 1. Introduction

Game theory is nowadays a well-established framework in theoretical computer science, enabling computer-aided design of computer systems that are correct-by-construction. It allows one to describe and analyse the possible interactions of antagonistic agents (or players) as in the controller synthesis problem, for instance. This problem asks, given a model of the environment of a

[^0]system, and of the possible actions of a controller, to compute a controller that constraints the environment to respect a given specification. Clearly, one cannot assume in general that the two players (the environment and the controller) lower value $\underline{\text { Val }}$ returns the greatest price that Max can ensure (letting the price be $+\infty$ in case the target locations are not reached).

An example of PTG is given in Figure 1, where the locations of Min and Max are represented by circles and rectangles respectively. The integers next to location. Moreover, there is only one clock $x$ in the game, which is never reset, and all guards on transitions are $x \in[0,1]$ (hence this guard is not displayed and transitions are only labeled by their respective discrete weight): this is an example of a simple priced timed game (we will define them properly later). configurations $(\ell, \nu)$ of the game, where $\ell$ is a location and $\nu$ is a real valuation of the clock in $[0,1]$. Let us comment on the optimal strategies for both players. From a configuration $\left(\ell_{4}, \nu\right)$, with $\nu \in[0,1]$, Max better waits until the clock takes value 1 , before taking the transition to $\ell_{f}$ (he is forced to move, by the
${ }_{50}$ rules of the game). Hence, Max's optimal value is $3(1-\nu)-7=-3 \nu-4$ from all configurations $\left(\ell_{4}, \nu\right)$. Symmetrically, it is easy to check that Min better waits as long as possible in $\ell_{7}$, hence his optimal value is $-16(1-\nu)$ from all configurations $\left(\ell_{7}, \nu\right)$. However, optimal value functions are not always that simple, see for instance the lower value function of $\ell_{1}$ on the bottom of Figure 1 ,
55 which is a piecewise affine function. To understand why value functions can be piecewise affine, consider the sub-game enclosed in the dotted rectangle in


Figure 1: A simple priced timed game (top) and the lower value function of location $\ell_{1}$ (bottom). Transitions without label have weight 0 .

Figure 1, and consider the value that Min can guarantee from a configuration of the form $\left(\ell_{3}, \nu\right)$ in this sub-game. Clearly, Min must decide how long he will spend in $\ell_{3}$ and whether he will go to $\ell_{4}$ or $\ell_{7}$. His optimal value from all $\left(\ell_{3}, \nu\right)$ is thus $\inf _{0 \leqslant t \leqslant 1-\nu} \min (4 t+3(1-(\nu+t))-7,4 t+6-16(1-(\nu+t)))=$ $\min (-3 \nu-4,16 \nu-10)$. Since $16 \nu-10 \geqslant-3 \nu-4$ if and only if $\nu \leqslant 6 / 19$, the best choice of Min is to move instantaneously to $\ell_{7}$ if $\nu \in[0,6 / 19]$ and to move instantaneously to $\ell_{4}$ if $\nu \in(6 / 19,1]$, hence the value function of $\ell_{3}$ (in the subgame) is a piecewise affine function with two pieces.

PTGs are a special case of hybrid games dAHM01, MPS95, WT97, that were independently investigated in BCFL04 and ABM04. For (non-necessarily turn-based) PTGs with non-negative prices, semi-algorithms are given to decide the value problem that is to say, whether the upper value of a location (the best price that Min can guarantee starting with a clock valuation 0 ), is below a given threshold. They have also shown that, under the strongly non-Zeno assumption on prices (asking the existence of $\kappa>0$ such that every cycle in the underlying region graph has a price at least $\kappa$ ), the proposed semi-algorithms always terminate. This assumption was justified in BBR05, BBM06] by showing that, without it, the existence problem, that is to decide whether Min has a strategy guaranteeing to reach a target location with a price below a given threshold, is indeed undecidable for PTGs with non-negative prices and three or more clocks. This result was recently extended in BJM14 to show that the value problem is also undecidable for PTGs with non-negative prices and four or more clocks. In [BCJ09, the undecidability of the existence problem has also been shown for PTGs with arbitrary price-rates (without prices on transitions), and two or more clocks. On a positive side, the value problem was shown decidable by [BLMR06] for PTGs with one clock when the prices are non-negative: a 3 -exponential time algorithm was first proposed, further refined in Rut11, HIJM13 into an exponential time algorithm. The key point of those algorithms is to reduce the
problem to the computation of optimal values in a restricted family of PTGs called Simple Priced Timed Games (SPTGs for short), where the underlying automata contain no guard, no reset, and the play is forced to stop after one time unit. More precisely, the PTG is decomposed into a sequence of SPTGs whose value functions are computed and re-assembled to yield the value function of the original PTG. Alternatively, and with radically different techniques, a pseudopolynomial time algorithm to solve one-clock PTGs with arbitrary prices on transitions, and price-rates restricted to two values amongst $\{-d, 0,+d\}$ (with $d \in \mathbb{N}$ ) was given in $\left[\mathrm{BGK}^{+} 14\right]$. A survey summarising those results was done in Bou15.

## Contributions.

Following the decidability results sketched above, we consider PTGs with one clock. We extend those results by considering arbitrary (positive and negative) prices. Indeed, all previous works on PTGs with only one clock (except $\left.\left.\mathrm{BGK}^{+} 14\right]\right)$ have considered non-negative weights only, and the status of the more general case with arbitrary weights has so far remained elusive. Yet, arbitrary weights are an important modeling feature. Consider, for instance, a system which can consume but also produce energy at different rates. In this case, energy consumption could be modeled as a positive price-rate, and production by a negative price-rate. In the untimed setting, such extension to negative weights has been considered in BGHM15, BGHM16: our result heavily builds upon techniques investigated in these works, as we will see later. Our main contribution is an exponential time algorithm to compute the value of one-clock SPTGs with arbitrary weights. While this result might sound limited due to the restricted class of simple PTGs we can handle, we recall that the previous works mentioned above BLMR06, Rut11, HIJM13] have demonstrated that solving SPTGs is a key result towards solving more general PTGs. Moreover, this algorithm is, as far as we know, the first to handle the full class of SPTGs with arbitrary weights, and we note that the solutions (either the algorithms or the proofs) known so far do not generalise to this case. Finally, as a side result, this algorithm allows us to solve the more general class of negative-reset-acyclic one-clock PTGs that we introduce. Thus, the whole class of PTGs with arbitrary weights and one clock remains open so far, our result may be seen as a potentially important milestone towards this goal.

## 2. Quantitative reachability games

The semantics of the priced timed games we study in this work can be expressed in the setting of quantitative reachability games as defined below. Intuitively, in such a game, two players (Min and Max) play by changing alternatively the current configuration of the game. The game ends when it reaches a final configuration, and Min has to pay a price associated with the sequence of configurations and of transitions taken (hence, he is trying to minimise this price).

Note that this framework of quantitative reachability games that we develop here (and for which we prove a determinacy result, see Theorem 1) can be applied to other settings than our priced timed games. For example, special cases of quantitative reachability games are finite quantitative reachability gameswhere the set of configurations is finite - that have been thoroughly studied in

BGHM16 under the name of min-cost reachability games. In this paper, we will rely on quantitative reachability games with uncountably many states as the semantics of priced timed games. Similarly, our quantitative reachability games could be used to fomalise the semantics of hybrid games [BBC06, BBJ+08] or any (non-probabilistic) game with a reachability objective.

We start our discussion by defining formally those games:
Definition 1 (Quantitative reachability games). $A$ quantitative reachability game is a tuple $G=\left(C=C_{\operatorname{Min}} \uplus C_{\mathrm{Max}}, \Sigma, E, F, p\right)$, where $C$ is the set of configurations (that does not need to be finite, nor even countable), partitioned into the set $C_{\mathrm{Min}}$ of configurations of player Min , and the set $C_{\mathrm{Max}}$ of configurations of player Max; $\Sigma$ is a (potentially infinite) alphabet whose elements are called letters; $E \subseteq C \times \Sigma \times C$ is the transition relation; $F \subseteq C$ is the set of final configurations; and $p:(C \times \Sigma)^{*} \times C \rightarrow \mathbb{R}$ maps each finite sequence $c_{1} a_{1} \cdots a_{n} c_{n+1}$ to a real number called the price of $c_{1} a_{1} \cdots a_{n} c_{n+1}$.

For the sake of exposure, we assume that there are no deadlocks in the game, i.e. for all configurations $c \in C$, there exists $c^{\prime} \in C$ and $a \in \Sigma$ such that $\left(c, a, c^{\prime}\right) \in E$. A finite play is a finite sequence $\rho=c_{1} a_{1} c_{2} \ldots c_{n}$ alternating between configurations and letters, and such that for all $i<n:\left(c_{i}, a_{i}, c_{i+1}\right) \in E$. In this case, we let $|\rho|=n$ be the length of the finite play. A play is an infinite sequence $\rho=c_{1} a_{1} c_{2} \cdots$ alternating between configurations and letters satisfying the same condition, i.e. for all $i \geqslant 1:\left(c_{i}, a_{i}, c_{i+1}\right) \in E$. In that case, we let $|\rho|$ be the least position $i$ such that $c_{i} \in F$, and $|\rho|=+\infty$ if there are no such positions. For the sake of clarity, we denote a play $c_{1} a_{1} c_{2} \cdots$ as $c_{1} \xrightarrow{a_{1}} c_{2} \cdots$, and similarly for finite plays.

We take the viewpoint of player Min who wants to reach a final configuration. Thus, the price of a play $\rho=c_{1} \xrightarrow{a_{1}} c_{2} \cdots$, denoted $\operatorname{Price}(\rho)$ is either $+\infty$ if $|\rho|=+\infty$ (this is the worst situation for Min, which explains why the price is maximal in this case); or $p\left(c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{n}\right)$ if $|\rho|=n$.

A strategy for player Min is a function $\sigma_{\text {Min }}$ mapping every finite play ending in a configuration $c \in C_{\text {Min }}$ to a transition $\left(c, a, c^{\prime}\right) \in E$. Strategies $\sigma_{\text {Max }}$ of player Max are defined accordingly. We let $\operatorname{Strat}_{\text {Min }}(G)$ and $\operatorname{Strat}_{\text {Max }}(G)$ be the sets of strategies of Min and Max, respectively. A pair $\left(\sigma_{\text {Min }}, \sigma_{\text {Max }}\right) \in \operatorname{Strat}_{\text {Min }}(\mathcal{G}) \times$ $\operatorname{Strat}_{\operatorname{Max}}(\mathcal{G})$ is called a profile of strategies. Together with an initial configuration $c_{1}$, it defines a unique play $\operatorname{Play}\left(c_{1}, \sigma_{\mathrm{Min}}, \sigma_{\mathrm{Max}}\right)=c_{1} \xrightarrow{a_{1}} c_{2} \cdots$ such that for all $i \geqslant 0:\left(c_{i}, a_{i}, c_{i+1}\right)=\sigma_{\operatorname{Min}}\left(c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{i}\right)$ if $c_{i} \in C_{\text {Min }} ;$ and $\left(c_{i}, a_{i}, c_{i+1}\right)=$ $\sigma_{\text {Max }}\left(c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{i}\right)$ if $c_{i} \in C_{\text {Max }}$. We let Play $\left(\sigma_{\text {Min }}\right)\left(\right.$ respectively, $\left.\operatorname{Play}\left(c_{1}, \sigma_{\text {Min }}\right)\right)$ be the set of plays that conform with $\sigma_{\text {Min }}$ (and start in $\left.c_{1}\right)$, and define $\operatorname{Play}\left(\sigma_{\text {Max }}\right)$ and Play $\left(c_{1}, \sigma_{\mathrm{Max}}\right)$ accordingly. Given an initial configuration $c_{1}$, the price of a strategy $\sigma_{\text {Min }}$ of Min is:

$$
\operatorname{Price}\left(c_{1}, \sigma_{\text {Min }}\right)=\sup _{\rho \in \operatorname{Play}\left(c_{1}, \sigma_{\text {Min }}\right)} \operatorname{Price}(\rho) .
$$

It matches the intuition to be the largest price that Min may pay while following strategy $\sigma_{\text {Min }}$. This definition is equal to $\sup _{\sigma_{\text {Max }}} \operatorname{Price}\left(\operatorname{Play}\left(c_{1}, \sigma_{\text {Min }}, \sigma_{\text {Max }}\right)\right)$, which is intuitively the highest price that Max can force Min to pay if Min follows $\sigma_{\mathrm{Min}}$. Similarly, given a strategy $\sigma_{\mathrm{Max}}$ of Max , we define the price of $\sigma_{\mathrm{Max}}$ as:

$$
\operatorname{Price}\left(c_{1}, \sigma_{\mathrm{Max}}\right)=\inf _{\rho \in \operatorname{Pay}\left(c_{1}, \sigma_{\mathrm{Max}}\right)} \operatorname{Price}(\rho)=\inf _{\sigma_{\mathrm{Min}}} \operatorname{Price}\left(\operatorname{Play}\left(c_{1}, \sigma_{\mathrm{Min}}, \sigma_{\mathrm{Max}}\right)\right) .
$$

It corresponds to the least price that Min can achieve once Max has fixed its strategy $\sigma_{\text {Max }}$.

From there, two different definitions of the value of a configuration $c_{1}$ arise, depending on which player chooses its strategy first. The upper value of $c_{1}$, defined as:

$$
\overline{\operatorname{Val}}\left(c_{1}\right)=\inf _{\sigma_{\mathrm{Min}}} \sigma_{\sigma_{\mathrm{Max}}} \operatorname{Pup} \operatorname{Price}\left(\operatorname{Play}\left(c_{1}, \sigma_{\mathrm{Min}}, \sigma_{\mathrm{Max}}\right)\right)
$$

corresponds to the least price that Min can ensure when choosing its strategy before Max, while the lower value, defined as:

$$
\underline{\operatorname{Val}}\left(c_{1}\right)=\sup _{\sigma_{\mathrm{Max}}} \inf _{\sigma_{\mathrm{Min}}} \operatorname{Price}\left(\operatorname{Play}\left(c_{1}, \sigma_{\mathrm{Min}}, \sigma_{\mathrm{Max}}\right)\right)
$$

corresponds to the least price that Min can ensure when choosing its strategy after Max. It is easy to see that $\underline{\mathrm{Val}}\left(c_{1}\right) \leqslant \overline{\mathrm{Val}}\left(c_{1}\right)$, which explains the chosen names. Indeed, if Min picks its strategy after Max, he has more information, and then can, in general, choose a better response.

In general, the order in which players choose their strategies can modify the outcome of the game. However, for quantitative reachability games, this makes no difference, and the value is the same whichever player picks his strategy first. This result, known as the determinacy property, is formalised here:

Theorem 1 (Determinacy of quantitative reachability games). For all quantitative reachability games $G$ and configurations $c_{1}, \overline{\operatorname{Val}}\left(c_{1}\right)=\underline{\mathrm{Val}}\left(c_{1}\right)$.
Proof. To establish this result, we rely on a general determinacy result of Donald Martin Mar75. This result concerns qualitative games (i.e. games where players either win or lose the game, and do not pay a price), called Gale-Stewart games. So, we first explain how to reduce a quantitative reachability game $G=(C=$ $\left.C_{\mathrm{Min}} \uplus C_{\mathrm{Max}}, \Sigma, E, F, p\right)$ to a family of such Gale-Stewart games $\operatorname{Threshold}(G, r)$ parametrised by a threshold $r \in \mathbb{R}$.

The Gale-Stewart game Threshold $(G, r)$ is played on an infinite tree whose vertices are owned by either of the players. A play is then a maximal branch in this tree, built as follows: the player who owns the root of the tree first picks a successor of the root that becomes the current vertex. Then, the player who owns this vertex gets to choose a successor that becomes the current one, etc. The game ends when a leaf is reached, where the winner is declared.

In our case, the vertices of $\operatorname{Threshold}(G, r)$ are the finite plays $c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{n}$ of $G$ starting from configuration $c_{1}$. Such a vertex $v=c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{n}$ is owned by Min iff $c_{n} \in C_{\text {Min }}$; otherwise $v$ belongs to Max. A vertex $v=c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{n}$ has successors iff $c_{n} \notin F$. In this case, the successors of $v$ are all the vertices $v \xrightarrow{a} c$ s.t. $\left(c_{n}, a, c\right) \in E$. Finally, a leaf $c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{n}$ (thus, with $\left.c_{n} \in F\right)$ is winning for Min iff $p\left(c_{1} \xrightarrow{a_{1}} c_{2} \cdots c_{n}\right) \leqslant r$.

As a consequence, the set of winning plays in $\operatorname{Threshold}(G, r)$ is:

$$
\text { Win }=\bigcup_{v \in L \text { s.t. } p(v) \leqslant r}\{\operatorname{branch}(v)\}
$$

where $L$ is the set of leaves of $\operatorname{Threshold}(G, r)$, and $\operatorname{branch}(v)$ is the (unique) branch from $c_{1}$ to $v$. Then,

$$
\operatorname{Win}=\bigcup_{v \in L \text { s.t. } p(v) \leqslant r} \operatorname{Cone}(v)
$$

where Cone $(v)$ is the set of plays in Threshold $(G, r)$ that visit $v$. Indeed, when $v$ is a leaf, the set Cone $(v)$ reduces to the singleton containing only branch $(v)$. Thus, the set of winning plays (for Min) is an open set, defined in the topology generated from the Cone $(v)$ sets, and we can apply Mar75 to conclude that Threshold $(G, r)$ is a determined game for all quantitative reachability games $G$ and all thresholds $r \in \mathbb{R}$ i.e. either Min or Max has a winning strategy from the root of the tree. Notice that strategies in $G$ and $\operatorname{Threshold}(G, r)$ are isomorphic.

We rely on this result to prove that $\underline{\operatorname{Val}}\left(c_{1}\right) \geqslant \overline{\operatorname{Val}}\left(c_{1}\right)$ in $G$ (the other inequality being always true). We consider two cases:

1. If $\overline{\operatorname{Val}}\left(c_{1}\right)=-\infty$, then $\underline{\operatorname{Val}}\left(c_{1}\right)$ being at most $\overline{\operatorname{Val}}\left(c_{1}\right)$ is $-\infty$ too.
2. If $\overline{\operatorname{Val}}\left(c_{1}\right)>-\infty$, consider any real number $t$ such that $t<\overline{\operatorname{Val}}\left(c_{1}\right)$. By definition of the upper value, for all strategies $\sigma_{\text {Min }}$, we have $\operatorname{Price}\left(c_{1}, \sigma_{\text {Min }}\right)>t$. Therefore, Min loses in the game Threshold ( $G, t$ ). By determinacy, Max wins in this game, i.e. there exists a strategy $\sigma_{\text {Max }}^{t}$ such that Price $\left(c_{1}, \sigma_{\text {Max }}^{t}\right)>$ $t$. By definition of the lower value, this ensures that $\operatorname{Val}\left(c_{1}\right) \geqslant t$. Therefore, $t<\overline{\operatorname{Val}}\left(c_{1}\right)$ implies $t \leqslant \underline{\operatorname{Val}}\left(c_{1}\right)$ : since this holds for all $t$, we have $\underline{\operatorname{Val}}\left(c_{1}\right) \leqslant \overline{\operatorname{Val}}\left(c_{1}\right)$.

In such determined games, we denote by Val the value of the game, defined as $\mathrm{Val}=\overline{\mathrm{Val}}=\underline{\mathrm{Val}}$.

## 3. Priced timed games

We are now ready to formally introduce the core model of our paper: priced timed games. We start by the formal definition, then study some properties of the value function of those games (Section 3.2). Next, we introduce the restricted class of simple priced timed games (Section 3.3) and close this section by discussing some special strategies (called switching strategies, see Section 3.4) that we will rely upon in our algorithms to solve priced timed games.

### 3.1. Notations and definitions.

Let $x$ denote a positive real-valued variable called clock. A guard (or clock constraint) is an interval with endpoints in $\mathbb{N} \cup\{+\infty\}$. We often abbreviate guards, writing for instance $x \leqslant 5$ instead of $[0,5]$. The set of all guards on the clock $x$ is called $\operatorname{Guard}(x)$. Let $S \subseteq \operatorname{Guard}(x)$ be a finite set of guards. We let $\llbracket S \rrbracket=\bigcup_{I \in S} I$. Assuming $M_{0}=0<M_{1}<\cdots<M_{k}$ are all the endpoints of the intervals in $S$ (to which we add 0 if needed), we let $\operatorname{Reg}_{S}=\left\{\left(M_{i}, M_{i+1}\right) \mid 0 \leqslant\right.$ $i \leqslant k-1\} \cup\left\{\left\{M_{i}\right\} \mid 0 \leqslant i \leqslant k\right\}$ be the set of regions of $S$. Observe that $\operatorname{Reg}_{S}$ is also a set of guards.

We rely on the notion of cost function to formalise the notion of optimal value function sketched in the introduction. Formally, for a set of guards $S \subseteq$ Guard $(x)$, a cost function over $S$ is a function $f: \llbracket \operatorname{Reg}_{S} \rrbracket \rightarrow \overline{\mathbb{R}}=\mathbb{R} \cup\{+\infty,-\infty\}$ such that over each region $r \in \operatorname{Reg}_{S}, f$ is either infinite or it is a continuous piecewise affine function with a finite set of cutpoints (points where the first derivative is not defined) $\left\{\kappa_{1}, \ldots, \kappa_{p}\right\} \subseteq \mathbb{Q}$, and satisfying $f\left(\kappa_{i}\right) \in \mathbb{Q}$ for all $1 \leqslant i \leqslant p$. In particular, if $f(r)=\{f(\nu) \mid \nu \in r\}$ contains $+\infty$ (respectively, $-\infty)$ for some region $r$, then $f(r)=\{+\infty\}(f(r)=\{-\infty\})$. We denote by $\mathrm{CF}_{S}$ the set of all cost functions over $S$.

In our algorithm to solve SPTGs, we will need to combine cost functions thanks to the $\triangleright$ operator. Let $f \in \mathrm{CF}_{S}$ and $f^{\prime} \in \mathrm{CF}_{S^{\prime}}$ be two cost functions on set of guards $S, S^{\prime} \subseteq \operatorname{Guard}(x)$, such that $\llbracket S \rrbracket \cap \llbracket S^{\prime} \rrbracket$ is a singleton. We let $f \triangleright f^{\prime}$ be the cost function in $\mathrm{CF}_{S \cup S^{\prime}}$ such that $\left(f \triangleright f^{\prime}\right)(\nu)=f(\nu)$ for all $\nu \in \llbracket \operatorname{Reg}_{S} \rrbracket$, and $\left(f \triangleright f^{\prime}\right)(\nu)=f^{\prime}(\nu)$ for all $\nu \in \llbracket \operatorname{Reg}_{S^{\prime}} \rrbracket \backslash \llbracket \operatorname{Reg}_{S} \rrbracket$. For example, let $S=\{\{0\},(0,1),\{1\}\}$ and $S^{\prime}=\{\{1\}\}$. We define the cost functions $f_{1}$ and $f_{2}$ such that $f_{1}$ is equal to $+\infty$ on the set of regions $\operatorname{Reg}_{S}$ and $f_{2}$ is equal to 0 on the set of regions $\operatorname{Reg}_{S^{\prime}}$. The cost function $f_{2} \triangleright f_{1} \in \mathrm{CF}_{S \cup S^{\prime}}$ is equal to $+\infty$ on $[0,1)$ and to 0 on $\{1\}$ and the cost function $f_{1} \triangleright f_{2} \in \mathrm{CF}_{S^{\prime}}$ is equal to $+\infty$ on $[0,1]$. Thus $f_{1} \triangleright f_{2}$ is equal to $f_{1}$ while $f_{2} \triangleright f_{1}$ extends $f_{2}$ with a $+\infty$ value on $[0,1)$.

We consider an extended notion of one-clock priced timed games (PTGs for short) allowing for the use of urgent locations, where only a zero delay can be spent, and final cost functions which are associated with all final locations and incur an extra price to be paid when ending the game in this location.

Definition 2. A priced timed game (PTG for short) $\mathcal{G}$ is a tuple $\left(L_{\mathrm{Min}}, L_{\mathrm{Max}}, L_{f}\right.$, $\left.L_{u}, \varphi, \Delta, \pi\right)$ where:

- $L_{\text {Min }}$ and $L_{\text {Max }}$ are finite sets of locations belonging respectively to player Min and Max. We assume $L_{\text {Min }} \cap L_{\text {Max }}=\emptyset$ and let $L=L_{\text {Min }} \cup L_{\text {Max }}$ be the set of all locations of the PTG;
- $L_{f} \subseteq L$ is a finite set of final locations;
- $L_{u} \subseteq L \backslash L_{f}$ is the set of urgent location $\mathbb{2}^{2}$;
- $\Delta \subseteq\left(L \backslash L_{f}\right) \times \operatorname{Guard}(x) \times\{\top, \perp\} \times L$ is a finite set of transitions. We denote by $S_{\mathcal{G}}=\left\{I \mid \exists \ell, R, \ell^{\prime}:\left(\ell, I, R, \ell^{\prime}\right) \in \Delta\right\}$ the set of all guards occurring on some transitions of the PTG;
- $\varphi=\left(\varphi_{\ell}\right)_{\ell \in L_{f}}$ associates to all locations $\ell \in L_{f} a$ final cost function, that is an affint ${ }^{3}$ cost function $\varphi_{\ell}$;
- $\pi: L \cup \Delta \rightarrow \mathbb{Z}$ is a mapping associating an integer weight (also called price) to all locations and transitions. In the case of a location $\ell$, we say that $\pi(\ell)$ is $\ell$ 's price-rate.

Intuitively, a transition $\left(\ell, I, R, \ell^{\prime}\right)$ changes the current location from $\ell$ to $\ell^{\prime}$ if the clock has value in $I$ and the clock is reset according to the Boolean $R$. We assume that, in all PTGs, the clock $x$ is bounded, i.e. there is $M \in \mathbb{N}$ such that for all guards $I \in S_{\mathcal{G}},\left.I \subseteq[0, M]\right|^{4}$ We denote by $\operatorname{Reg}_{\mathcal{G}}$ the set $\operatorname{Reg}_{S_{\mathcal{G}}}$ of regions of $\mathcal{G}$. We further denot ${ }^{5}$ by $\Pi_{\mathcal{G}}^{\mathrm{tr}}$, $\Pi_{\mathcal{G}}^{\text {loc }}$ and $\Pi_{\mathcal{G}}^{\mathrm{fin}}$ respectively the values $\max _{\delta \in \Delta}|\pi(\delta)|$, $\max _{\ell \in L}|\pi(\ell)|$ and $\sup _{\nu \in[0, M]} \max _{\ell \in L}\left|\varphi_{\ell}(\nu)\right|=\max _{\ell \in L} \max \left(\left|\varphi_{\ell}(0)\right|,\left|\varphi_{\ell}(M)\right|\right)$

[^1](the last equality holds because we have assumed that $\varphi_{\ell}$ is affine). That is, $\Pi_{\mathcal{G}}^{\mathrm{tr}}$, $\Pi_{\mathcal{G}}^{\text {loc }}$ and $\Pi_{\mathcal{G}}^{\text {fin }}$ are the largest absolute values of the transition prices, location prices and final cost functions.

As announced in the first section, the semantics of a PTG $\mathcal{G}=\left(L_{\text {Min }}, L_{\text {Max }}, L_{f}\right.$, $\left.L_{u}, \boldsymbol{\varphi}, \Delta, \pi\right)$ is given by a quantitative reachability game

$$
G_{\mathcal{G}}=\left(\operatorname{Conf}_{\mathcal{G}}, \Sigma=\left(\mathbb{R}^{+} \times \Delta \times \mathbb{R}\right), E, F=\left(L_{f} \times \mathbb{R}^{+}\right), p\right)
$$

that we describe now. Note that, from now on, we often confuse the PTG $\mathcal{G}$ with its semantics $G_{\mathcal{G}}$, writing, for instance 'the configurations of $\mathcal{G}$ ' instead of: 'the configurations of $G_{\mathcal{G}}$ '. We also lift the Price, $\overline{\mathrm{Val}}, \underline{\mathrm{Val}}$ and Val functions, and the notions of plays from $G_{\mathcal{G}}$ to $\mathcal{G}$. A configuration of $\mathcal{G}$ is a pair $s=(\ell, \nu) \in L \times \mathbb{R}^{+}$, where $\ell$ and $\nu$ are respectively the current location and clock value of $\mathcal{G}$. We denote by $\mathrm{Conf}_{\mathcal{G}}$ the set of all configurations of $\mathcal{G}$. Let $(\ell, \nu)$ and $\left(\ell^{\prime}, \nu^{\prime}\right)$ be two configurations, let $\delta=\left(\ell, I, R, \ell^{\prime}\right) \in \Delta$ be a transition of $\mathcal{G}$ and $t \in \mathbb{R}^{+}$be a delay. Then, $\left((\ell, \nu),(t, \delta, c),\left(\ell^{\prime}, \nu^{\prime}\right)\right) \in E$, iff:
(i) $\ell \in L_{u}$ implies $t=0$ (no time can elapse in urgent locations);
(ii) $\nu+t \in I$ (the guard is satisfied);
(iii) $R=\top$ implies $\nu^{\prime}=0$ (when the clock is reset);
(iv) $R=\perp$ implies $\nu^{\prime}=\nu+t$ (when the clock is not reset);
(v) $c=\pi(\delta)+t \times \pi(\ell)$ (the price of $(t, \delta)$ takes into account the price-rate of $\ell$, the delay spent in $\ell$, and the price of $\delta$ ).
In this case, we say that there is a $(t, \delta)$-transition from $(\ell, \nu)$ to $\left(\ell^{\prime}, \nu^{\prime}\right)$ with price $c$, and we denote this by $(\ell, \nu) \xrightarrow{t, \delta, c}\left(\ell^{\prime}, \nu^{\prime}\right)$. For two configurations $s$ and $s^{\prime}$, we also write $s \xrightarrow{c} s^{\prime}$ whenever there are $t$ and $\delta$ such that $s \xrightarrow{t, \delta, c} s^{\prime}$. Observe that, since the alphabet of $G_{\mathcal{G}}$ is $\mathbb{R}^{+} \times \Delta \times \mathbb{R}$, and its set of configurations is Conf $_{\mathcal{G}}$, plays of $\mathcal{G}$ are of the form $\rho=\left(\ell_{1}, \nu_{1}\right) \xrightarrow{t_{1}, \delta_{1}, c_{1}}\left(\ell_{2}, \nu_{2}\right) \cdots$. Finally, the price function $p$ is obtained by summing the price of the play (transitions and time spent in the locations) and the final cost function if applicable. Formally, for a finite play $\rho=\left(\ell_{1}, \nu_{1}\right) \xrightarrow{t_{1}, \delta_{1}, c_{1}}\left(\ell_{2}, \nu_{2}\right) \cdots\left(\ell_{n}, \nu_{n}\right)$ with $\forall k<n, \ell_{k} \notin L_{f}$, if $\ell_{n} \in L_{f}$ then $p(\rho)=\sum_{i=1}^{n-1} c_{i}+\varphi_{\ell_{|\rho|}}\left(\nu_{n}\right)$ else $p(\rho)=\sum_{i=1}^{n-1} c_{i}$.

As sketched in the introduction, we consider optimal reachability-price games on PTGs, where the aim of player Min is to reach a location of $L_{f}$ while minimising the price. Since the semantics of PTGs is defined in terms of quantitative reachability games, we can apply Theorem 1, and deduce that all PTGs $\mathcal{G}$ are determined. Hence, for all PTGs the value function Val is well-defined, and we denote it by $\mathrm{Val}_{\mathcal{G}}$ when we need to emphasise the game it refers to.

For example, consider the PTG on the top of Figure 1. Using the final cost function $\varphi$ constantly equal to 0 , its value function for location $\ell_{1}$ is represented on the right. The play $\rho=\left(\ell_{1}, 0\right) \xrightarrow{0, t_{1,2}, 0}\left(\ell_{2}, 0\right) \xrightarrow{1 / 4, t_{2,3},-3.5}\left(\ell_{3}, 1 / 4\right) \xrightarrow{0, t_{3,7}, 6}$ $\left(\ell_{7}, 1 / 4\right) \xrightarrow{3 / 4, t_{7, f},-12}\left(\ell_{f}, 1\right)$ where $t_{n, m}=\left(\ell_{n},[0,1], \perp, \ell_{m}\right)$ ends in the unique final location $\ell_{f}$ and its price is $p(\rho)=0-3.5+6-12=-9.5$.

Let us fix a PTG $\mathcal{G}$ with initial configuration $c_{1}$. We say that a strategy $\sigma_{\text {Min }}$ of Min is optimal if Price $\left(c_{1}, \sigma_{\mathrm{Min}}\right)=\operatorname{Val}_{\mathcal{G}}\left(c_{1}\right)$, i.e., it ensures Min to enforce the value of the game, whatever Max does. Similarly, $\sigma_{\text {Min }}$ is $\varepsilon$-optimal, for $\varepsilon>0$, if $\operatorname{Price}\left(c_{1}, \sigma_{\text {Min }}\right) \leqslant \operatorname{Val}_{\mathcal{G}}\left(c_{1}\right)+\varepsilon$. And, symmetrically, a strategy $\sigma_{\text {Max }}$ of Max is optimal (respectively, $\varepsilon$-optimal) if $\operatorname{Price}\left(c_{1}, \sigma_{\mathrm{Max}}\right)=\mathrm{Va}_{\mathcal{G}}\left(c_{1}\right)$ (respectively, $\left.\operatorname{Price}\left(c_{1}, \sigma_{\mathrm{Max}}\right) \geqslant \operatorname{Val}_{\mathcal{G}}\left(c_{1}\right)+\varepsilon\right)$. the function such that $\operatorname{Va}(\ell)(\nu)=V^{\prime}(\ell, \nu)$ for all $\nu \in \mathbb{R}^{+}$. Then, we show that, for all $\ell \operatorname{Va}_{( }(\ell)$ is a piecewise continuous function that might exhibit discontinuities only on the borders of the regions of $\mathrm{Reg}_{\mathcal{G}}$.

Theorem 2. For all (one-clock) PTGs $\mathcal{G}$, for all $r \in \operatorname{Reg}_{\mathcal{G}}$, for all $\ell \in L$, $\mathrm{Val}_{\mathcal{G}}(\ell)$ is either infinite or continuous over $r$.

Proof. Our goal is to show that for every location $\ell$, region $r \in \operatorname{Reg}_{\mathcal{G}}$ and valuations $\nu$ and $\nu^{\prime}$ in $r$,

$$
\left|\operatorname{Val}(\ell, \nu)-\operatorname{Val}\left(\ell, \nu^{\prime}\right)\right| \leqslant \Pi^{\mathrm{loc}}\left|\nu-\nu^{\prime}\right| .
$$

This is equivalent to showing:

$$
\operatorname{Val}(\ell, \nu) \leqslant \operatorname{Val}\left(\ell, \nu^{\prime}\right)+\Pi^{\mathrm{loc}}\left|\nu-\nu^{\prime}\right| \quad \text { and } \quad \operatorname{Val}\left(\ell, \nu^{\prime}\right) \leqslant \operatorname{Val}(\ell, \nu)+\Pi^{\mathrm{loc}}\left|\nu-\nu^{\prime}\right| .
$$

As those two equations are symmetric with respect to $\nu$ and $\nu^{\prime}$, we only have to show either of them. We will thus focus on the latter, which, by using the upper value, can be reformulated as: for all strategies $\sigma_{\text {Min }}$ of Min, there exists a strategy $\sigma_{\text {Min }}^{\prime}$ such that Price $\left(\left(\ell, \nu^{\prime}\right), \sigma_{\text {Min }}^{\prime}\right) \leqslant \operatorname{Price}\left((\ell, \nu), \sigma_{\text {Min }}\right)+\Pi^{\text {loc }}\left|\nu-\nu^{\prime}\right|$. Note that this last equation is equivalent to say that there exists a function $g$ mapping plays $\rho^{\prime}$ from $\left(\ell, \nu^{\prime}\right)$, consistent with $\sigma_{\text {Min }}^{\prime}$ (i.e. such that $\rho^{\prime}=\operatorname{Play}\left(\left(\ell, \nu^{\prime}\right), \sigma_{\text {Min }}^{\prime}, \sigma_{\text {Max }}\right)$ for some strategy $\sigma_{\mathrm{Max}}$ of Max ) to plays from $(\ell, \nu)$, consistent with $\sigma_{\mathrm{Min}}$, such that:

$$
\operatorname{Price}\left(\rho^{\prime}\right) \leqslant \operatorname{Price}\left(g\left(\rho^{\prime}\right)\right)+\Pi^{\mathrm{loc}}\left|\nu-\nu^{\prime}\right| .
$$

Let $r \in \operatorname{Reg}_{\mathcal{G}}, \nu, \nu^{\prime} \in r$ and $\sigma_{\text {Min }}$ be a strategy of Min. We define $\sigma_{\text {Min }}^{\prime}$ and $g$ by induction on the length of the finite play that is given as argument; more precisely, we define $\sigma_{\text {Min }}^{\prime}\left(\rho_{1}^{\prime}\right)$ and $g\left(\rho_{2}^{\prime}\right)$ by induction on $k$, for all plays $\rho_{1}^{\prime}$ and $\rho_{2}^{\prime}$ from $\left(\ell, \nu^{\prime}\right)$, consistent with $\sigma_{\text {Min }}^{\prime}$ of length $k-1$ and $k$, respectively. We also show during this induction that for each play $\rho^{\prime}=\left(\ell_{1}, \nu_{1}^{\prime}\right) \xrightarrow{c_{1}^{\prime}} \cdots \xrightarrow{c_{k-1}^{\prime}}\left(\ell_{k}, \nu_{k}^{\prime}\right)$ from $\left(\ell, \nu^{\prime}\right)$, consistent with $\sigma_{\text {Min }}^{\prime}$, if we let $\left(\ell_{1}, \nu_{1}\right) \xrightarrow{c_{1}} \cdots \xrightarrow{c_{\ell-1}}\left(\ell_{\ell}, \nu_{\ell}\right)=g\left(\rho^{\prime}\right)$ :
(i) $\rho^{\prime}$ and $g\left(\rho^{\prime}\right)$ have the same length, i.e. $|\rho|=\ell=k=\left|\rho^{\prime}\right|$,
(ii) for every $i \in\{1, \ldots, k\}, \nu_{i}$ and $\nu_{i}^{\prime}$ are in the same region, i.e. there exists a region $r^{\prime} \in \operatorname{Reg}_{\mathcal{G}}$ such that $\nu_{i} \in r^{\prime}$ and $\nu_{i}^{\prime} \in r^{\prime}$,
(iii) $\left|\nu_{k}-\nu_{k}^{\prime}\right| \leqslant\left|\nu-\nu^{\prime}\right|$,
(iv) $\operatorname{Price}\left(\rho^{\prime}\right) \leqslant \operatorname{Price}\left(g\left(\rho^{\prime}\right)\right)+\Pi^{\mathrm{loc}}\left(\left|\nu-\nu^{\prime}\right|-\left|\nu_{k}-\nu_{k}^{\prime}\right|\right)$.

Notice that no property is required on the strategy $\sigma_{\text {Min }}^{\prime}$ for finite plays that do not start in $\left(\ell, \nu^{\prime}\right)$.

If $k=0, \sigma_{\text {Min }}^{\prime}$ does not have to be defined. Moreover, in that case, $\rho^{\prime}=\left(\ell, \nu^{\prime}\right)$ and $g\left(\rho^{\prime}\right)=(\ell, \nu)$. Both plays have length $0, \nu$ and $\nu^{\prime}$ are in the same region by hypothesis of the induction, and Price $\left(\rho^{\prime}\right)=\operatorname{Price}\left(g\left(\rho^{\prime}\right)\right)=0$, therefore all four properties are true.

Let us suppose now that the construction is done for a given $k \geqslant 1$, and perform it for $k+1$. We start with the construction of $\sigma_{\text {Min }}^{\prime}$. To that extent,


Figure 2: The definition of $t^{\prime}$ when (a) $\nu_{k}^{\prime} \leqslant \nu_{k}$, (b) $\nu_{k}<\nu_{k}^{\prime}<\nu_{k}+t$, (c) $\nu_{k}<\nu_{k}+t<\nu_{k}^{\prime}$.
consider a play $\rho^{\prime}=\left(\ell_{1}, \nu_{1}^{\prime}\right) \xrightarrow{c_{1}^{\prime}} \cdots \xrightarrow{c_{k-1}^{\prime}}\left(\ell_{k}, \nu_{k}^{\prime}\right)$ from $\left(\ell, \nu^{\prime}\right)$, consistent with

By induction hypothesis $\left|\tilde{\rho}^{\prime}\right|=\left|g\left(\tilde{\rho}^{\prime}\right)\right|$, thus: i holds, i.e. $\left|\rho^{\prime}\right|=\left|g\left(\rho^{\prime}\right)\right|$. Moreover, $\nu_{k+1}$ and $\nu_{k+1}^{\prime}$ are also in the same region as either they are equal to $\nu_{k}+t$ and $\nu_{k}^{\prime}+t^{\prime}$, respectively, or $\delta$ contains a reset in which case $\nu_{k+1}=\nu_{k+1}^{\prime}=0$ which proves ii To prove iiii notice that we always have either $\nu_{k}+t=\nu_{k}^{\prime}+t^{\prime}$ or $\nu_{k} \leqslant \nu_{k}+t \leqslant \nu_{k}^{\prime}=\nu_{k}^{\prime}+t$ or $\nu_{k}^{\prime} \leqslant \nu_{k}^{\prime}+t \leqslant \nu_{k}=\nu_{k}+t$. In all of these possibilities, we have $\left|\left(\nu_{k}+t\right)-\left(\nu_{k}^{\prime}+t^{\prime}\right)\right| \leqslant\left|\nu_{k}-\nu_{k}^{\prime}\right|$. We finally check property iv In both cases:

$$
\begin{aligned}
\operatorname{Price}\left(\rho^{\prime}\right) & =\operatorname{Price}\left(\tilde{\rho}^{\prime}\right)+\pi(\delta)+t^{\prime} \pi\left(\ell_{k}\right) \\
& \leqslant \operatorname{Price}\left(g\left(\tilde{\rho}^{\prime}\right)\right)+\Pi^{\mathrm{loc}}\left(\left|\nu-\nu^{\prime}\right|-\left|\nu_{k}-\nu_{k}^{\prime}\right|\right)+\pi(\delta)+t^{\prime} \pi\left(\ell_{k}\right) \\
& =\operatorname{Price}\left(g\left(\rho^{\prime}\right)\right)+\left(t^{\prime}-t\right) \pi\left(\ell_{k}\right)+\Pi^{\mathrm{loc}}\left(\left|\nu-\nu^{\prime}\right|-\left|\nu_{k}-\nu_{k}^{\prime}\right|\right) .
\end{aligned}
$$

If $\delta$ contains no reset, let us prove that

$$
\begin{equation*}
\left|t^{\prime}-t\right|=\left|\nu_{k}-\nu_{k}^{\prime}\right|-\left|\nu_{k+1}^{\prime}-\nu_{k+1}\right| . \tag{1}
\end{equation*}
$$

Indeed, since $t^{\prime}=\nu_{k+1}^{\prime}-\nu_{k}^{\prime}$ and $t=\nu_{k+1}-\nu_{k}$, we have $\left|t^{\prime}-t\right|=\mid \nu_{k+1}^{\prime}-\nu_{k}^{\prime}-$ $\left(\nu_{k+1}-\nu_{k}\right) \mid$. Then, two cases are possible: either $t^{\prime}=\max \left(0, \nu_{k}+t-\nu_{k}^{\prime}\right)$ or $t=\max \left(0, \nu_{k}^{\prime}+t^{\prime}-\nu_{k}\right)$. So we have three different possibilities:

- if $t^{\prime}+\nu_{k}^{\prime}=t+\nu_{k}, \nu_{k+1}^{\prime}=\nu_{k+1}$, thus $\left|t^{\prime}-t\right|=\left|\nu_{k}-\nu_{k}^{\prime}\right|=\left|\nu_{k}-\nu_{k}^{\prime}\right|-$ $\left|\nu_{k+1}^{\prime}-\nu_{k+1}\right|$.
- if $t=0$, then $\nu_{k}=\nu_{k+1} \geqslant \nu_{k+1}^{\prime} \geqslant \nu_{k}^{\prime}$, thus $\left|\nu_{k+1}^{\prime}-\nu_{k}^{\prime}-\left(\nu_{k+1}-\nu_{k}\right)\right|=$ $\nu_{k+1}^{\prime}-\nu_{k}^{\prime}=\left(\nu_{k}-\nu_{k}^{\prime}\right)-\left(\nu_{k}-\nu_{k+1}^{\prime}\right)=\left|\nu_{k}-\nu_{k}^{\prime}\right|-\left|\nu_{k+1}^{\prime}-\nu_{k+1}\right|$.
- if $t^{\prime}=0$, then $\nu_{k}^{\prime}=\nu_{k+1}^{\prime} \geqslant \nu_{k+1} \geqslant \nu_{k}$, thus $\left|\nu_{k+1}^{\prime}-\nu_{k}^{\prime}-\left(\nu_{k+1}-\nu_{k}\right)\right|=$ $\nu_{k+1}-\nu_{k}=\left(\nu_{k}^{\prime}-\nu_{k}\right)-\left(\nu_{k}^{\prime}-\nu_{k+1}\right)=\left|\nu_{k}-\nu_{k}^{\prime}\right|-\left|\nu_{k+1}^{\prime}-\nu_{k+1}\right|$.

If $\delta$ contains a reset, then $\nu_{k+1}^{\prime}=\nu_{k+1}$. If $t^{\prime}=\nu_{k}+t-\nu_{k}^{\prime}$, we have that $\left|t^{\prime}-t\right|=\left|\nu_{k}-\nu_{k}^{\prime}\right|$. Otherwise, either $t=0$ and $t^{\prime} \leqslant \nu_{k}-\nu_{k}^{\prime}$, or $t^{\prime}=0$ and $t \leqslant \nu_{k}^{\prime}-\nu_{k}$.

In all cases, we have proved (1). Coupled with the fact that $\left|P\left(\ell_{k}\right)\right| \leqslant \Pi^{\text {loc }}$, we conclude that:

$$
\operatorname{Price}\left(\rho^{\prime}\right) \leqslant \operatorname{Price}\left(g\left(\rho^{\prime}\right)\right)+\Pi^{\mathrm{loc}}\left(\left|\nu-\nu^{\prime}\right|-\left|\nu_{k+1}-\nu_{k+1}^{\prime}\right|\right)
$$

Now that $\sigma_{\text {Min }}^{\prime}$ and $g$ are defined (noticing that $g$ is stable by prefix, we extend naturally its definition to infinite plays), notice that for all plays $\rho^{\prime}$ from $\left(\ell, \nu^{\prime}\right)$ consistent with $\sigma_{\text {Min }}^{\prime}$, either $\rho^{\prime}$ does not reach a final location and its price is $+\infty$, but in this case $g\left(\rho^{\prime}\right)$ has also price $+\infty$; or $\rho^{\prime}$ is finite. In this case let $\nu_{k}^{\prime}$ be the clock valuation of its last configuration, and $\nu_{k}$ be the clock valuation of the last configuration of $g\left(\rho^{\prime}\right)$. Combining (iii) and (iv) we have $\operatorname{Price}\left(\rho^{\prime}\right) \leqslant \operatorname{Price}\left(g\left(\rho^{\prime}\right)\right)+\Pi^{\text {loc }}\left|\nu-\nu^{\prime}\right|$ which concludes the proof.

Remark 1. Let us consider the example in Figure 3 (that we describe informally since we did not properly define games with multiple clocks), with clocks $x$ and $y$. One can easily check that, starting from a configuration ( $\ell_{0}, 0,0.5$ ) in location $\ell_{0}$ and where $x=0$ and $y=0.5$, the following cycle can be taken: $\left(\ell_{0}, 0,0.5\right) \xrightarrow{0, \delta_{0}, 0}$ $\left(\ell_{1}, 0,0.5\right) \xrightarrow{0.5, \delta_{1}, 2.5}\left(\ell_{2}, 0.5,0\right) \xrightarrow{0.5, \delta_{2},-2.5}\left(\ell_{0}, 0,0.5\right)$, where $\delta_{0}, \delta_{1}$ and $\delta_{2}$ denote respectively the transitions from $\ell_{0}$ to $\ell_{1}$; from $\ell_{1}$ to $\ell_{2}$; and from $\ell_{2}$ to $\ell_{0}$. Observe that the price of this cycle is null, and that no other delays can be played, hence $\overline{\operatorname{Val}}\left(\ell_{0}, 0,0.5\right)=0$. However, starting from a configuration $\left(\ell_{0}, 0,0.6\right)$, and following the same path, yields the cycle $\left(\ell_{0}, 0,0.6\right) \xrightarrow{0, e_{0}, 0}\left(\ell_{1}, 0,0.6\right) \xrightarrow{0.4, e_{1}, 2}$ $\left(\ell_{2}, 0.4,0\right) \xrightarrow{0.6, e_{2},-3}\left(\ell_{0}, 0,0.6\right)$ with price -1 . Hence, $\overline{\operatorname{Val}}\left(\ell_{0}, 0,0.6\right)=-\infty$, and 390 the function is not continuous although both valuations $(0,0.5)$ and $(0,0.6)$ are in the same region. Observe that this holds even for priced timed automata, since our example requires only one player.


Figure 3: A PTG with 2 clocks whose value function is not continuous inside a region.

### 3.3. Simple priced timed games.

As sketched in the introduction, our main contribution is to solve the special case of simple one-clock priced timed games with arbitrary weights. Formally, an $r$-SPTG, with $r \in \mathbb{Q}^{+} \cap[0,1]$, is a PTG $\mathcal{G}=\left(L_{\text {Min }}, L_{\text {Max }}, L_{f}, L_{u}, \boldsymbol{\varphi}, \Delta, \pi\right)$ such that for all transitions $\left(\ell, I, R, \ell^{\prime}\right) \in \Delta, I=[0, r]$ (the clock is also bounded by $r)$ and $R=\perp$. Hence, transitions of $r$-SPTGs are henceforth denoted by $\left(\ell, \ell^{\prime}\right)$, dropping the guard and the reset. Then, an SPTG is a 1-SPTG. This paper is mainly devoted to prove the following result on SPTGs.

Theorem 3. Let $\mathcal{G}$ be an SPTG. Then, for all locations $\ell \in L$, the function $\operatorname{Val}(\ell)$ is either infinite, or continuous and piecewise-affine with at most an exponential number of cutpoints. The value functions for all locations, as well as a pair of optimal strategies $\left(\sigma_{\mathrm{Min}}, \sigma_{\mathrm{Max}}\right)$ (that always exist if no values are infinite) can be computed in exponential time.

### 3.4. Switching strategies.

Let us now discuss a class of (simple) strategies that are sufficient to play optimally. Those strategies, called switching strategies, will be instrumental in proving the result above. Roughly speaking, Max has always a memoryless optimal strategy, while Min might need (finite) memory to play optimally-it is already the case in untimed quantitative reachability games with arbitrary weights [BGHM16]. Moreover, these strategies are finitely representable (recall that even a memoryless strategy depends on the current configuration and that there are infinitely many in our time setting).

We start by formalising Max's strategies, thanks to the notion of finite positional strategies:

Definition 3 (FP-strategies). A strategy $\sigma$ is a finite positional strategy ( $F P$ strategy for short) iff it is a memoryless strategy (i.e. for all finite plays $\rho_{1}=$ $\rho_{1}^{\prime} \xrightarrow{c_{1}} s$ and $\rho_{2}=\rho_{2}^{\prime} \xrightarrow{c_{2}} s$ ending in the same configuration, we have $\sigma\left(\rho_{1}\right)=$ $\left.\sigma\left(\rho_{2}\right)\right)$ and for all locations $\ell$, there exists a finite sequence of rationals $0 \leqslant \nu_{1}^{\ell}<$ $\nu_{2}^{\ell}<\cdots<\nu_{k}^{\ell}=1$ and a finite sequence of transitions $\delta_{1}, \ldots, \delta_{k} \in \Delta$ such that
(i) for all $1 \leqslant i \leqslant k$, for all $\nu \in\left(\nu_{i-1}^{\ell}, \nu_{i}^{\ell}\right]$, either $\sigma(\ell, \nu)=\left(0, \delta_{i}\right)$, or $\sigma(\ell, \nu)=$ $\left(\nu_{i}^{\ell}-\nu, \delta_{i}\right)\left(\right.$ assuming $\left.\nu_{0}^{\ell}=\min \left(0, \nu_{1}^{\ell}\right)\right) ;$ and
(ii) if $\nu_{1}^{\ell}>0$, then $\sigma(\ell, 0)=\left(\nu_{1}^{\ell}, \delta_{1}\right)$.

We let $\operatorname{pts}(\sigma)$ be the set of $\nu_{i}^{\ell}$ for all $\ell$ and $i$, and $\operatorname{int}(\sigma)$ be the set of all successive intervals generated by $\operatorname{pts}(\sigma)$. Finally, we let $|\sigma|=|\operatorname{int}(\sigma)|$ be the size of $\sigma$. Intuitively, in an interval $\left(\nu_{i-1}^{\ell}, \nu_{i}^{\ell}\right], \sigma$ always returns the same move: either to take immediately $\delta_{i}$ or to wait until the clock reaches the endpoint $\nu_{i}^{\ell}$ and then take $\delta_{i}$.

While Max's strategies can be memoryless, we observe that Min may require memory to play optimally as shown in the following example taken from BGHM16. Consider the SPTG of Figure 4, where $W$ is a positive integer, and every location has price-rate 0 (thus, it is an untimed game, as originally studied). We claim that the values of locations $\ell_{1}$ and $\ell_{2}$ are both $-W$. Indeed, consider the following strategy for Min: during each of the first $W$ visits to $\ell_{2}$ (if any), go to $\ell_{1}$; else, go to $\ell_{f}$. Clearly, this strategy ensures that the final location $\ell_{f}$ will eventually be reached, and that either
(i) transition $\left(\ell_{1}, \ell_{3}\right)$ (with weight $-W$ ) will eventually be traversed; or


Figure 4: An SPTG where Min needs memory to play optimally
(ii) transition $\left(\ell_{1}, \ell_{2}\right)$ (with weight -1 ) will be traversed at least $W$ times.

Hence, in all plays following this strategy, the price will be at most $-W$. This strategy allows Min to secure $-W$, but he cannot ensure a lower price, since Max always has the opportunity to take the transition $\left(\ell_{1}, \ell_{f}\right)$ (with weight $-W$ ) instead of cycling between $\ell_{1}$ and $\ell_{2}$. Hence, Max's optimal choice is to follow the transition $\left(\ell_{1}, \ell_{f}\right)$ as soon as $\ell_{1}$ is reached, securing a price of $-W$. The runs $\rho=\left(\ell_{1}, \nu\right) \xrightarrow{c_{1}} \cdots \xrightarrow{c_{k-1}}\left(\ell_{k}, \nu^{\prime}\right) \in \operatorname{Play}\left(\sigma_{\mathrm{Min}}\right)$ with $\ell_{1}=\ell_{k}$, and $\nu, \nu^{\prime}$ in the same interval of $\operatorname{int}\left(\sigma_{\mathrm{Min}}\right)$, the sum of prices of discrete transitions is at most -1 , i.e. $\pi\left(\ell_{1}, \ell_{2}\right)+\cdots+\pi\left(\ell_{k-1}, \ell_{k}\right) \leqslant-1$. Notice that all finite plays $\rho \in \operatorname{Play}\left(\sigma_{\text {Min }}\right)$ with all clock valuations in the same interval $I$ of $\operatorname{int}\left(\sigma_{\text {Min }}\right)$ verify Price $(\rho) \leqslant|I| \Pi^{\text {loc }}+|L| \Pi^{\text {tr }}-|\rho| /|L|$. Indeed, the price of $\rho$ is the sum of the price generated by staying in locations, which is bounded by $|I| \Pi^{\text {loc }}$, and the price of the transitions. One can extract at least $|\rho| /|L|$ cycles with transition prices at most - 1 (by definition of an NC-strategy), and what remains is of size at most $|L|$, ensuring that the transition price is bounded by $|L| \Pi^{\text {tr }}-|\rho| /|L|$.

Then, by splitting runs among intervals of int $\left(\sigma_{\text {Min }}\right)$, we can easily obtain that all finite plays $\rho \in \operatorname{Play}\left(\sigma_{\text {Min }}\right)$ verify Price $(\rho) \leqslant \Pi^{\text {loc }}+\left(2\left|\sigma_{\text {Min }}\right|-1\right) \times|L| \Pi^{\text {tr }}-(|\rho|-$ $\left.\left|\sigma_{\text {Min }}\right|\right) /|L|$. Indeed, letting $I_{1}, I_{2}, \ldots, I_{k}$ the interval of int $\left(\sigma_{\text {Min }}\right)$ visited during $\rho$ (with $k \leqslant\left|\sigma_{\text {Min }}\right|$ ), one can split $\rho$ into $k$ runs $\rho=\rho_{1} \xrightarrow{c_{1}} \rho_{2} \xrightarrow{c_{2}} \cdots \rho_{k}$ such that in
$\rho_{i}$ all clock values are in $I_{i}$ (remember that SPTGs contain no reset transitions). By the previous inequality, we have $\operatorname{Price}\left(\rho_{i}\right) \leqslant\left|I_{i}\right| \Pi^{\text {loc }}+|L| \Pi^{\mathrm{tr}}-\left|\rho_{i}\right| /|L|$. Thus, also splitting prices $c_{i}$ with respect to discrete price and price of delaying, we obtain Price $(\rho)=\sum_{i=1}^{k} \operatorname{Price}\left(\rho_{i}\right)+\sum_{i=1}^{k-1} c_{i} \leqslant\left(2\left|\sigma_{\text {Min }}\right|-1\right) \times|L| \Pi^{\text {tr }}+\Pi^{\mathrm{loc}}-$ $\left(|\rho|-\left|\sigma_{\text {Min }}\right|\right) / \mid L$, since $|\rho| \leqslant \sum_{i}\left|\rho_{i}\right|+k \leqslant \sum_{i}\left|\rho_{i}\right|+\left|\sigma_{\text {Min }}\right|$ and $\sum_{i}\left|I_{i}\right| \leqslant 1$.

To characterise the fact that $\sigma_{\text {Min }}$ must allow Min to reach a price which is small enough, without necessarily reaching a target state, we define the fake value of an NC-strategy $\sigma_{\text {Min }}$ from a configuration $s$ as fake $\mathcal{G}_{\mathcal{G}}^{\sigma_{\text {Min }}}(s)=\sup \{\operatorname{Price}(\rho) \mid$ $\rho \in \operatorname{Play}\left(s, \sigma_{\text {Min }}\right), \rho$ reaches a target $\}$, i.e. the value obtained when ignoring the $\sigma_{\text {Min }}$-induced plays that do not reach the target. Thus, clearly, fake $\mathcal{G}_{\mathcal{G}}^{\sigma_{\text {Min }}}(s) \leqslant$ $\mathrm{Val}^{\sigma_{\mathrm{Min}}}(s)$. We say that an NC-strategy is fake-optimal if its fake value, in every configuration, is equal to the optimal value of the configuration in the game. This is justified by the following result whose proof relies on the switching strategies described before:

Lemma 4. If $\operatorname{Val}_{\mathcal{G}}(\ell, \nu) \neq+\infty$, for all $\ell$ and $\nu$, then for all $N C$-strategies $\sigma_{\text {Min }}$, there is a strategy $\sigma_{\text {Min }}^{\prime}$ such that $\mathrm{Val}_{\mathcal{G}}^{\sigma_{\text {Min }}^{\prime}}(s) \leqslant$ fake $_{\mathcal{G}}^{\sigma_{\text {Min }}}(s)$ for all configurations s. In particular, if $\sigma_{\mathrm{Min}}$ is a fake-optimal NC-strategy, then $\sigma_{\text {Min }}^{\prime}$ is an optimal (switching) strategy of the SPTG.

Proof. We suppose known an attractor strategy for Min: it exists thanks to the hypothesis on the finiteness of the values. From every configuration, it reaches a final location with a price bounded above by a given constant $M$. Notice first that, under the hypothesis that no configurations of the SPTG have value $-\infty$, we have fake ${ }_{\mathcal{G}}^{\sigma_{\text {Min }}}(s)>-\infty$ for a configuration $s$. Otherwise, consider the strategy $\sigma_{\text {Min }}^{\prime}$ obtained by playing $\sigma_{\text {Min }}$ until having computed a price bounded above by a fixed integer $N \in \mathbb{Z}$, in which case we switch to the attractor strategy. By the previous inequality, the switch is sure to happen since the right term tends to $-\infty$ when the length of $\rho$ tends to $\infty$. Then, we know that the value guaranteed by $\sigma_{\text {Min }}^{\prime}$ is at most $N$, implying that the optimal value $\operatorname{Val}(s)$ is $-\infty$, which contradicts the hypothesis. Then, to prove the result of the lemma, consider the strategy $\sigma_{\text {Min }}^{\prime}$ obtained by playing $\sigma_{\text {Min }}$ until having computed a price bounded above by the finite value fake $\mathcal{G}_{\mathcal{G}}^{\sigma_{\operatorname{Min}}}(s)-M$, in which case we switch to the attractor strategy. Once again, the switch is sure to happen, implying that every play conforming to $\sigma_{\text {Min }}$ reaches the target: moreover, the price of such a play is necessarily at most fake ${ }^{\sigma_{\text {Min }}}(s)$ by construction. Then, we directly obtain that $\operatorname{Val}_{\mathcal{G}}^{\sigma_{\text {Min }}^{\prime}}(s) \leqslant$ fake $_{\mathcal{G}}^{\sigma_{\text {Min }}}(s)$.

Then, an SPTG is called finitely optimal if
(i) Min has a fake-optimal NC-strategy;
(ii) Max has an optimal FP-strategy; and
(iii) $\operatorname{Val}_{\mathcal{G}}(\ell)$ is a cost function, for all locations $\ell$.

The central point in establishing Theorem 3 will thus be to prove that all SPTGs are finitely optimal, as this guarantees the existence of well-behaved optimal strategies and value functions. We will also show that they can be computed in exponential time. The proof is by induction on the number of urgent locations of the SPTG. In Section 4, we address the base case of SPTGs with urgent locations only (where no time can elapse). Since these SPTGs are very close to the untimed min-cost reachability games of [BGHM16], we adapt $\operatorname{Val}(\ell, \nu)$, so that $\mathrm{Val}_{\nu}$ is the vector we want to compute. Since no time can elapse, it consists in an adaptation of the techniques developed in BGHM16 to solve (untimed) min-cost reachability games. The main difference concerns the prices being rational (and not integers) and the presence of final cost functions

Following the arguments of BGHM16, we first observe that locations $\ell$ with values $\operatorname{Val}_{\nu}(\ell)=+\infty$ and $\operatorname{Val}_{\nu}(\ell)=-\infty$ can be pre-computed (using respectively attractor and mean-payoff techniques) and removed from the game without changing the other values. Then, because of the particular structure of the game $\mathcal{G}$ (where a real price is paid only on the target location, all other prices being integers), for all plays $\rho$, $\operatorname{Price}(\rho)$ is a value from the set $\mathbb{Z}_{\nu, \varphi}=$ $\mathbb{Z}+\left\{\varphi_{\ell}(\nu) \mid \ell \in L_{f}\right\}$. We further define $\mathbb{Z}_{\nu, \varphi}^{+\infty}=\mathbb{Z}_{\nu, \varphi} \cup\{+\infty\}$. Clearly, $\mathbb{Z}_{\nu, \varphi}$ contains at most $\left|L_{f}\right|$ values between two consecutive integers, i.e.

$$
\begin{equation*}
\forall i \in \mathbb{Z} \quad\left|[i, i+1] \cap \mathbb{Z}_{\nu, \varphi}\right| \leqslant\left|L_{f}\right| \tag{2}
\end{equation*}
$$

Then, we define an operator $\mathcal{F}:\left(\mathbb{Z}_{\nu, \varphi}^{+\infty}\right)^{L} \rightarrow\left(\mathbb{Z}_{\nu, \boldsymbol{\varphi}}^{+\infty}\right)^{L}$ mapping every vector $\boldsymbol{x}=\left(x_{\ell}\right)_{\ell \in L}$ of $\left(\mathbb{Z}_{\nu, \boldsymbol{\varphi}}^{+\infty}\right)^{L}$ to $\mathcal{F}(\boldsymbol{x})=\left(\mathcal{F}(\boldsymbol{x})_{\ell}\right)_{\ell \in L}$ defined by

$$
\mathcal{F}(\boldsymbol{x})_{\ell}= \begin{cases}\varphi_{\ell}(\nu) & \text { if } \ell \in L_{f} \\ \max _{\left(\ell, \ell^{\prime}\right) \in \Delta}\left(\pi\left(\ell, \ell^{\prime}\right)+x_{\ell^{\prime}}\right) & \text { if } \ell \in L_{\mathrm{Max}} \\ \min _{\left(\ell, \ell^{\prime}\right) \in \Delta}\left(\pi\left(\ell, \ell^{\prime}\right)+x_{\ell^{\prime}}\right) & \text { if } \ell \in L_{\mathrm{Min}} .\end{cases}
$$

We will obtain $\mathrm{Val}_{\nu}$ as the limit of the sequence $\left(\boldsymbol{x}^{(i)}\right)_{i \geqslant 0}$ defined by $x_{\ell}^{(0)}=+\infty$ if $\ell \notin L_{f}$, and $x_{\ell}^{(0)}=\varphi_{\ell}(\nu)$ if $\ell \in L_{f}$, and then $\boldsymbol{x}^{(i)}=\mathcal{F}\left(\boldsymbol{x}^{(i-1)}\right)$ for $i \geqslant 0$.

The intuition behind is that $\boldsymbol{x}^{(i)}$ is the value of the game (when the clock takes valuation $\nu$ ) if we impose that Min must reach the target within $i$ steps (and pays a price of $+\infty$ if it fails to do so). Formally, for a play $\rho=\ell_{0} \ell_{1} \cdots$, we let Price ${ }^{\leqslant i}(\rho)=\operatorname{Price}(\rho)$ if $\ell_{k} \in L_{f}$ for some $k \leqslant i$, and Price ${ }^{\leqslant i}(\rho)=+\infty$ otherwise. We further let

$$
\overline{\operatorname{Va}}_{\nu}^{\leqslant i}(\ell)=\inf _{\sigma_{\text {Min }}} \sup _{\sigma_{\mathrm{Max}}} \operatorname{Price}{ }^{\leqslant i}\left(\operatorname{Play}\left((\ell, \nu), \sigma_{\mathrm{Max}}, \sigma_{\mathrm{Min}}\right)\right)
$$

where $\sigma_{\text {Min }}$ and $\sigma_{\text {Max }}$ are respectively strategies of Min and Max. Lemma 6 of [BGHM16] allows us to easily obtain that:

Lemma 5. For all $i \geqslant 0$, and $\ell \in L: \boldsymbol{x}_{\ell}^{(i)}=\overline{\mathrm{Val}}_{\nu}^{\leqslant i}(\ell)$.
Sketch of proof. This is proved by induction on $i$. It is trivial for $i=0$, and playing one more step amounts to computing one more iterate of $\mathcal{F}$.

Now, let us study how the sequence $\left(\overline{\mathrm{Val}}_{\nu}{ }^{\leqslant i}\right)_{i \geqslant 0}$ behaves and converges to the finite values of the game. Using again the same arguments as in BGHM16 (in particular, that $\mathcal{F}$ is a monotonic and Scott-continuous operator over the complete lattice $\left.\left(\mathbb{Z}_{\nu, \varphi}^{+\infty}\right)^{L}\right)$, the sequence $\left(\overline{\mathrm{VaI}}_{\nu}{ }_{\nu}^{i i}\right)_{i \geqslant 0}$ converges towards the greatest fixed point of $\mathcal{F}$. Let us now show that $\mathrm{Val}_{\nu}$ is actually this greatest fixed point. First, Corollary 8 of [BGHM16] can be adapted to obtain

Lemma 6. For all $\ell \in L: \overline{\mathrm{Va}}_{\nu}^{\leqslant|L|}(\ell) \leqslant|L| \Pi^{\mathrm{tr}}+\Pi^{\mathrm{fin}}$.
Proof. Denoting by $\operatorname{Attr}_{i}(S)$ the $i$-steps attractor of set $S$, and assuming that $\operatorname{Attr}_{-1}(S)=\emptyset$ for all $S$, we can establish by induction on $j$ that: for all locations $0 \leqslant j \leqslant|L|:$
(i) $j<k$ implies $\overline{\mathrm{Val}}_{\nu}^{\leqslant j}(\ell)=+\infty$ and
(ii) $j \geqslant k$ implies $\overline{\mathrm{Val}}_{\nu}^{\leqslant j}(\ell) \leqslant j W+\Pi^{\mathrm{fin}}$ and $\overline{\mathrm{Val}}_{\nu}^{\leqslant j}(\ell) \in \mathbb{Z}_{\nu, \boldsymbol{\varphi}}$.

Then, the result is obtained by taking $j=|L|$ in (ii).
The next step is to show that the values that can be computed along the sequence (still assuming that $\operatorname{Val}(\ell, \nu)$ is finite for all $\ell$ ) are taken from a finite set:

Lemma 7. For all $i \geqslant 0$ and for all $\ell \in L$ :
where PossVal ${ }_{\nu}$ has cardinality bounded by $\left|L_{f}\right| \times\left((2|L|-1) \Pi^{\mathrm{tr}}+2 \Pi^{\mathrm{fin}}+1\right)$.
Proof. Following the proof of BGHM16, Lemma 9], it is easy to show that if Min can secure, from some vertex $\ell$, a price less than $-(|L|-1) \Pi^{\mathrm{tr}}-\Pi^{\mathrm{fin}}$, i.e. $\operatorname{Val}(\ell, \nu)<-(|L|-1) \Pi^{\mathrm{tr}}-\Pi^{\mathrm{fin}}$, then it can secure an arbitrarily small price from that configuration, i.e. $\operatorname{Val}(\ell, \nu)=-\infty$, which contradicts our hypothesis that the value is finite.

```
Algorithm 1: solveInstant \((\mathcal{G}, \nu)\)
    Input: \(r\)-SPTG \(\mathcal{G}=\left(L_{\text {Min }}, L_{\text {Max }}, L_{f}, L_{u}, \boldsymbol{\varphi}, \Delta, \pi\right)\), a valuation \(\nu \in[0, r]\)
    foreach \(\ell \in L\) do
        if \(\ell \in L_{f}\) then \(\mathrm{X}(\ell):=\varphi_{\ell}(\nu)\) else \(\mathrm{X}(\ell):=+\infty\)
    repeat
        \(\mathrm{X}_{\text {pre }}:=\mathrm{X}\)
        foreach \(\ell \in L_{\text {Max }}\) do \(\mathrm{X}(\ell):=\max _{\left(\ell, \ell^{\prime}\right) \in \Delta}\left(\pi\left(\ell, \ell^{\prime}\right)+\mathrm{X}_{\text {pre }}\left(\ell^{\prime}\right)\right)\)
        foreach \(\ell \in L_{\text {Min }}\) do \(X(\ell):=\min _{\left(\ell, \ell^{\prime}\right) \in \Delta}\left(\pi\left(\ell, \ell^{\prime}\right)+\mathrm{X}_{\text {pre }}\left(\ell^{\prime}\right)\right)\)
        foreach \(\ell \in L\) such that \(X(\ell)<-(|L|-1) \Pi^{\operatorname{tr}}-\Pi^{\mathrm{fin}}\) do \(X(\ell):=-\infty\)
    until \(\mathrm{X}=\mathrm{X}_{\text {pre }}\)
    return \(X\)
```

Hence, for all $i \geqslant 0$, for all $\ell: \overline{\operatorname{Val}}_{\nu}^{\leqslant i}(\ell) \geqslant \operatorname{Val}(\ell, \nu)>-(|L|-1) \Pi^{\operatorname{tr}}-\Pi^{\mathrm{fin}}$. By Lemma 6 and since the sequence is non-increasing, we conclude that, for all $i \geqslant 0$ and for all $\ell \in L$ :

$$
-(|L|-1) \Pi^{\operatorname{tr}}-\Pi^{\mathrm{fin}}<\overline{\mathrm{Va}}_{\nu}^{\leqslant|L|+i}(\ell) \leqslant|L| \Pi^{\operatorname{tr}}+\Pi^{\mathrm{fin}}
$$

Since all $\overline{\mathrm{Va}}_{\nu}^{\leqslant|L|+i}(\ell)$ are also in $\mathbb{Z}_{\nu, \boldsymbol{\varphi}}$, we conclude that $\overline{\mathrm{Val}}_{\nu}^{\leqslant|L|+i}(\ell) \in \operatorname{PossVal}_{\nu}$ for all $i \geqslant 0$. The upper bound on the size of $\mathrm{Poss}^{\mathrm{Val}}{ }_{\nu}$ is established by (2).

This allows us to bound the number of iterations needed for the sequence to stabilise. The worst case is when all locations are assigned a value bounded below by $-(|L|-1) \Pi^{\operatorname{tr}}-\Pi^{\text {fin }}$ from the highest possible values where all vertices are assigned a value bounded above by $|L| \Pi^{\mathrm{tr}}+\Pi^{\mathrm{fin}}$, which is itself reached after $|L|$ steps. Hence:

Corollary 8. The sequence $\left(\overline{\mathrm{Val}}_{\nu}{ }_{\nu}\right)_{i \geqslant 0}$ stabilises after a number of steps at most $\left|L_{f}\right| \times|L| \times\left((2|L|-1) \Pi^{\mathrm{tr}}+2 \Pi^{\mathrm{fin}}+1\right)+|L|$.

Next, the proofs of [BGHM16, Lemma 10 and Corollary 11] allow us to conclude that this sequence converges towards the value $\mathrm{Val}_{\nu}$ of the game (when all values are finite), which proves that the value iteration scheme of Algorithm 1 computes exactly $\mathrm{Val}_{\nu}$ for all $\nu \in[0, r]$. Indeed, this algorithm also works when some values are not finite. As a corollary, we obtain a characterisation of the possible values of $\mathcal{G}$ :

Corollary 9. For all $r$-SPTGs $\mathcal{G}$ with only urgent locations, for all locations $\ell \in$
$L$ and valuations $\nu \in[0, r], \operatorname{Val}(\ell, \nu)$ is contained in the set $\mathrm{PossVal}_{\nu} \cup\{-\infty,+\infty\}$ of cardinal polynomial in $|L|, \Pi^{\mathrm{tr}}$, and $\Pi^{\mathrm{fin}}$, i.e. pseudo-polynomial with respect to the size of $\mathcal{G}$.

Finally, Sections 3.4 of BGHM16 explain how to compute simultaneously optimal strategies for both players. In our context, this allows us to obtain for every valuation $\nu \in[0, r]$ and location $\ell$ of an $r$-SPTG, such that $\operatorname{Val}(\ell, \nu) \notin$ $\{-\infty,+\infty\}$, an optimal FP-strategy for Max, and an optimal switching strategy for Min.


Figure 5: Network of affine functions defined by $F_{\mathcal{G}}$ : functions in bold are final affine functions of $\mathcal{G}$, whereas non-bold ones are their translations with weights $k \in\left[-(|L|-1) \Pi^{\operatorname{tr}},|L| \Pi^{\operatorname{tr}}\right] \cap \mathbb{Z}$. Poss $\mathrm{CP}_{\mathcal{G}}$ is the set of abscisses of intersections points, represented by black disks.

### 4.2. Study of the complete value functions: $\mathcal{G}$ is finitely optimal

Now let us explain how we can reduce the computation of $\operatorname{Val}_{\mathcal{G}}(\ell): \nu \in$ $[0, r] \mapsto \operatorname{Val}(\ell, \nu)$ (for all $\ell$ ) to a finite number of calls to solveInstant. We first study a precise characterisation of these functions, in particular showing that these are cost functions of $\mathrm{CF}_{\{[0, r]\}}$.

We first define the set $\mathrm{F}_{\mathcal{G}}$ of affine functions over $[0, r]$ as follows:

$$
\mathrm{F}_{\mathcal{G}}=\left\{k+\varphi_{\ell} \mid \ell \in L_{f} \wedge k \in\left[-(|L|-1) \Pi^{\mathrm{tr}},|L| \Pi^{\operatorname{tr}}\right] \cap \mathbb{Z}\right\}
$$

Observe that this set is finite and that its cardinality is $2|L|^{2} \Pi^{\text {tr }}$, pseudopolynomial in the size of $\mathcal{G}$. Moreover, as a direct consequence of Corollary 9 , this set contains enough information to compute the value of the game in each possible valuation of the clock, in the following sense:

Lemma 10. For all $\ell \in L$, for all $\nu \in[0, r]:$ if $\operatorname{Val}(\ell, \nu)$ is finite, then there is $f \in \mathrm{~F}_{\mathcal{G}}$ such that $\operatorname{Val}(\ell, \nu)=f(\nu)$.

Using the continuity of $\mathrm{Val}_{\mathcal{G}}$ (Theorem 2), this shows that all the cutpoints of $\mathrm{Val}_{\mathcal{G}}$ are intersections of functions from $\mathrm{F}_{\mathcal{G}}$, i.e. belong to the set of possible cutpoints

$$
\text { PossCP } \mathcal{G}=\left\{\nu \in[0, r] \mid \exists f_{1}, f_{2} \in \mathrm{~F}_{\mathcal{G}} \quad f_{1} \neq f_{2} \wedge f_{1}(\nu)=f_{2}(\nu)\right\}
$$

This set is depicted in Figure 5 on an example. Observe that Poss $\mathrm{CP}_{\mathcal{G}}$ contains at most $\left|\mathrm{F}_{\mathcal{G}}\right|^{2}=4\left|L_{f}\right|^{4}\left(\Pi^{\mathrm{tr}}\right)^{2}$ points (also a pseudo-polynomial in the size of $\mathcal{G}$ ) since all functions in $F_{\mathcal{G}}$ are affine, and can thus intersect at most once with every other function. Moreover, Poss $C_{\mathcal{G}} \subseteq \mathbb{Q}$, since all functions of $\mathrm{F}_{\mathcal{G}}$ take rational values in 0 and $r \in \mathbb{Q}$. Thus, for all $\ell, \mathrm{Va}_{\mathcal{G}}(\ell)$ is a cost function (with cutpoints in Poss $\mathrm{CP}_{\mathcal{G}}$ and pieces from $\left.\mathrm{F}_{\mathcal{G}}\right)$. Since $\mathrm{Val}_{\mathcal{G}}(\ell)$ is a piecewise affine function, we can characterise it completely by computing only its value on its
cutpoints. Hence, we can reconstruct $\mathrm{Val}_{\mathcal{G}}(\ell)$ by calling solveInstant on each rational valuation $\nu \in \operatorname{Poss} \mathrm{CP}_{\mathcal{G}}$. From the optimal strategies computed along solveInstant, we can also reconstruct a fake-optimal NC-strategy for Min and an optimal FP-strategy for Max, hence: $r$-SPTGs with only urgent locations.

Let us conclude the proof that $r$-SPTGs with only urgent locations are finitely optimal by showing that Min has a fake-optimal NC-strategy, and Max has an optimal FP-strategy. Let $\nu_{1}, \nu_{2}, \ldots, \nu_{k}$ be the sequence of elements from $\operatorname{Poss} \mathrm{CP}_{\mathcal{G}}$ in increasing order, and let us assume $\nu_{0}=0$. For all $0 \leqslant i \leqslant k$ let $f_{i}^{\ell}$ be the function from $\mathrm{F}_{\mathcal{G}}$ that defines the piece of $\operatorname{Va|_{\mathcal {G}}(\ell )\text {intheinterval}[\nu _{i-1},\nu _{i}]}$ (we have shown above that such an $f_{i}^{\ell}$ always exists). Formally, for all $0 \leqslant i \leqslant k$, $f_{i}^{\ell} \in \mathrm{F}_{\mathcal{G}}$ verifies $\operatorname{Val}(\ell, \nu)=f_{i}^{\ell}(\nu)$, for all $\nu \in\left[\nu_{i-1}^{\ell}, \nu_{i}^{\ell}\right]$. Next, for all $1 \leqslant i \leqslant k$, let $\mu_{i}$ be a value taken in the middle of $\left[\nu_{i-1}, \nu_{i}\right]$, i.e. $\mu_{i}=\frac{\nu_{i}+\nu_{i-1}}{2}$. Note that all $\mu_{i}$ 's are rational values since all $\nu_{i}$ 's are. By applying solveInstant in each $\mu_{i}$, we can compute $\left(\mathrm{Val}_{\mathcal{G}}\left(\ell, \mu_{i}\right)\right)_{\ell \in L}$, and we can extract an optimal memoryless strategy $\sigma_{\text {Max }}^{i}$ for Max and an optimal switching strategy $\sigma_{\text {Min }}^{i}$ for Min. Thus we know that, for all $\ell \in L$, playing $\sigma_{\text {Min }}^{i}$ (respectively, $\sigma_{\mathrm{Max}}^{i}$ ) from $\left(\ell, \mu_{i}\right)$ allows Min (respectively, Max) to ensure a price at most (respectively, at least) $\operatorname{Val}_{\mathcal{G}}\left(\ell, \mu_{i}\right)=f_{i}^{\ell}\left(\mu_{i}\right)$. However, it is easy to check that the bound given by $f_{i}^{\ell}\left(\mu_{i}\right)$ holds in every valuation, i.e. for all $\ell$, for all $\nu$

$$
\operatorname{Price}\left((\ell, \nu), \sigma_{\operatorname{Min}}^{i}\right) \leqslant f_{i}^{\ell}(\nu) \quad \text { and } \quad \operatorname{Price}\left((\ell, \nu), \sigma_{\operatorname{Max}}^{i}\right) \geqslant f_{i}^{\ell}(\nu) .
$$

This holds because:
(i) Min can play $\sigma_{\text {Min }}^{i}$ from all clock valuations (in $[0, r]$ ) since we are considering an $r$-SPTG; and
(ii) Max does not have more possible strategies from an arbitrary valuation $\nu \in[0, r]$ than from $\mu_{i}$, because all locations are urgent and time cannot elapse (neither from $\nu$, nor from $\mu_{i}$ ).

And symmetrically for Max.
We conclude that Min can consistently play the same strategy $\sigma_{\text {Min }}^{i}$ from all configurations $(\ell, \nu)$ with $\nu \in\left[\nu_{i-1}, \nu_{i}\right]$ and secure a price which is at most $f_{i}^{\ell}(\nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)$, i.e. $\sigma_{\text {Min }}^{i}$ is optimal on this interval. By definition of $\sigma_{\text {Min }}^{i}$, it is
easy to extract from it a fake-optimal NC-strategy (actually, $\sigma_{\text {Min }}^{i}$ is a switching strategy described by a pair ( $\sigma_{\text {Min }}^{1}, \sigma_{\text {Min }}^{2}$ ), and $\sigma_{\text {Min }}^{1}$ can be used to obtain the fake-optimal NC-strategy). The same reasoning applies to strategies of Max and we conclude that Max has an optimal FP-strategy.

## 5. Finite optimality of general SPTGs

In this section, we consider SPTGs with non-urgent locations. We first prove that all such SPTGs are finitely optimal. Then, we introduce Algorithm 2 to compute optimal values and strategies of SPTGs. Throughout the section, we fix an SPTG $\mathcal{G}=\left(L_{\text {Min }}, L_{\text {Max }}, L_{f}, L_{u}, \varphi, \Delta, \pi\right)$ with non-urgent locations. Before presenting our core contributions, let us explain how we can detect locations with infinite values. As already argued, we can compute $\operatorname{Val}(\ell, 1)$ for all $\ell$ assuming all locations are urgent, since time cannot elapse anymore when the clock has valuation 1. This can be done with solveInstant. Then, by continuity, $\operatorname{Val}(\ell, 1)=+\infty($ respectively, $\operatorname{Val}(\ell, 1)=-\infty)$ if and only if $\operatorname{Val}(\ell, \nu)=+\infty$ (respectively, $\operatorname{Val}(\ell, \nu)=-\infty)$ for all $\nu \in[0,1]$. We remove from the game all locations with infinite value without changing the values of other locations. Thus, we henceforth assume that $\operatorname{Val}(\ell, \nu) \in \mathbb{R}$ for all $(\ell, \nu)$.

### 5.1. The $\mathcal{G}_{L^{\prime}, r}$ construction.

To prove finite optimality of SPTGs and to establish correctness of our algorithm, we rely in both cases on a construction that consists in decomposing $\mathcal{G}$ into a sequence of SPTGs with more urgent locations. Intuitively, a game with more urgent locations is easier to solve since it is closer to an untimed game (in particular, when all locations are urgent, we can apply the techniques of Section (4). More precisely, given a set $L^{\prime}$ of non-urgent locations, and a valuation $r_{0} \in[0,1]$, we will define a (possibly infinite) sequence of valuations $1=r_{0}>$ $r_{1}>\cdots$ and a sequence $\mathcal{G}_{L^{\prime}, r_{0}}, \mathcal{G}_{L^{\prime}, r_{1}}, \ldots$ of SPTGs such that
(i) all locations of $\mathcal{G}$ are also present in each $\mathcal{G}_{L^{\prime}, r_{i}}$, except that the locations of $L^{\prime}$ are now urgent; and
(ii) for all $i \geqslant 0$, the value function of $\mathcal{G}_{L^{\prime}, r_{i}}$ is equal to $\mathrm{Val}_{\mathcal{G}}$ on the interval $\left[r_{i+1}, r_{i}\right]$. Hence, we can re-construct $\mathrm{Val}_{\mathcal{G}}$ by assembling well-chosen parts of the values functions of the $\mathcal{G}_{L^{\prime}, r_{i}}$ (assuming $\inf _{i} r_{i}=0$ ).

This basic result will be exploited in two directions. First, we prove by induction on the number of urgent locations that all SPTGs are finitely optimal, by reconstructing $\mathrm{Val}_{\mathcal{G}}$ (as well as optimal strategies) as a $\triangleright$-concatenation of the value functions of a finite sequence of SPTGs with one more urgent locations. The base case, with only urgent locations, is solved by Proposition 11. This construction suggests a recursive algorithm in the spirit of [BLMR06, Rut11] (for non-negative prices). Second, we show that this recursion can be avoided (see Algorithm 22. Instead of turning locations urgent one at a time, this algorithm makes them all urgent and computes directly the sequence of SPTGs with only urgent locations. Its proof of correctness relies on the finite optimality of SPTGs and, again, on our basic result linking the values functions of $\mathcal{G}$ and games $\mathcal{G}_{L^{\prime}, r_{i}}$.

Let us formalise these constructions. Let $\mathcal{G}$ be an SPTG, let $r \in[0,1]$ be an endpoint, and let $\boldsymbol{x}=\left(x_{\ell}\right)_{\ell \in L}$ be a vector of rational values. Then, wait $(\mathcal{G}, r, \boldsymbol{x})$ is an $r$-SPTG in which both players may now decide, in all nonurgent locations $\ell$, to wait until the clock takes value $r$, and then to stop


Figure 6: The condition (3) (in the case $L^{\prime}=\emptyset$ and $\ell \in L_{\text {Min }}$ ): graphically, it means that the slope between every two points of the plot in $[a, r]$ (represented with a thick line) is greater than or equal to $-\pi(\ell)$ (represented with dashed line).
the game, adding the price $x_{\ell}$ to the current price of the play. Formally, wait $(\mathcal{G}, r, \boldsymbol{x})=\left(L_{\mathrm{Min}}, L_{\mathrm{Max}}, L_{f}^{\prime}, L_{u}, \varphi^{\prime}, T^{\prime}, \pi^{\prime}\right)$ is such that

- $L_{f}^{\prime}=L_{f} \uplus\left\{\ell^{f} \mid \ell \in L \backslash L_{u}\right\} ;$
- for all $\ell^{\prime} \in L_{f}$ and $\nu \in[0, r], \varphi_{\ell^{\prime}}^{\prime}(\nu)=\varphi_{\ell^{\prime}}(\nu)$, for all $\ell \in L \backslash L_{u}, \varphi_{\ell^{f}}^{\prime}(\nu)=$ $(r-\nu) \cdot \pi(\ell)+x_{\ell} ;$
- $T^{\prime}=T \cup\left\{\left(\ell,[0, r], \perp, \ell^{f}\right) \mid \ell \in L \backslash L_{u}\right\}$;
- for all $\delta \in T^{\prime}, \pi^{\prime}(\delta)=\pi(\delta)$ if $\delta \in T$, and $\pi^{\prime}(\delta)=0$ otherwise.

Then, we let $\mathcal{G}_{r}=\operatorname{wait}\left(\mathcal{G}, r,\left(\operatorname{Val}_{\mathcal{G}}(\ell, r)\right)_{\ell \in L}\right)$, i.e. the game obtained thanks to wait by letting $\boldsymbol{x}$ be the value of $\mathcal{G}$ in $r$. This first transformation does not alter the value of the game, for valuations before $r$ :

Lemma 12. For all $\nu \in[0, r]$ and locations $\ell, \operatorname{Val}_{\mathcal{G}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}_{r}}(\ell, \nu)$.
Next, we make locations urgent. For a set $L^{\prime} \subseteq L \backslash L_{u}$ of non-urgent locations, we let $\mathcal{G}_{L^{\prime}, r}$ be the SPTG obtained from $\mathcal{G}_{r}$ by making urgent every location $\ell$ of $L^{\prime}$. Observe that, although all locations $\ell \in L^{\prime}$ are now urgent in $\mathcal{G}_{L^{\prime}, r}$, their clones $\ell^{f}$ allow the players to wait until $r$. When $L^{\prime}$ is a singleton $\{\ell\}$, we write $\mathcal{G}_{\ell, r}$ instead of $\mathcal{G}_{\{\ell\}, r}$.

While the construction of $\mathcal{G}_{r}$ does not change the value of the game, introducing urgent locations does. Yet, we can characterise an interval [a,r] on which the value functions of $\mathcal{H}=\mathcal{G}_{L^{\prime}, r}$ and $\mathcal{H}^{+}=\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$ coincide, as stated by the next proposition. The interval $[a, r]$ depends on the slopes of the pieces of $\mathrm{Val}_{\mathcal{H}^{+}}$as depicted in Figure 6 for each location $\ell$ of Min, the slopes of the pieces of $\mathrm{Val}_{\mathcal{H}^{+}}$contained in $[a, r]$ should be $\leqslant-\pi(\ell)$ (and $\geqslant-\pi(\ell)$ when $\ell$
belongs to Max). It is proved by lifting optimal strategies of $\mathcal{H}^{+}$into $\mathcal{H}$, and the slope of $\operatorname{Val}_{\mathcal{G}}(\ell)$ in-between $\nu$ and $\nu^{\prime}$ by $\operatorname{slope}_{\mathcal{G}}^{\ell}\left(\nu, \nu^{\prime}\right)$, formally defined by $\operatorname{slope}_{\mathcal{G}}^{\ell}\left(\nu, \nu^{\prime}\right)=\frac{\operatorname{Val}_{\mathcal{G}}\left(\ell, \nu^{\prime}\right)-\operatorname{Val}_{\mathcal{G}}(\ell, \nu)}{\nu^{\prime}-\nu}$.
Proposition 13. Let $0 \leqslant a<r \leqslant 1, L^{\prime} \subseteq L \backslash L_{u}$ and $\ell \notin L^{\prime} \cup L_{u}$ a non-urgent location of Min (respectively, Max). Assume that $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$ is finitely optimal, and for all $a \leqslant \nu_{1}<\nu_{2} \leqslant r$

$$
\begin{equation*}
\operatorname{slope}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}^{\ell}\left(\nu_{1}, \nu_{2}\right) \geqslant-\pi(\ell) \quad(\text { respectively }, \leqslant-\pi(\ell)) \tag{3}
\end{equation*}
$$

Then, for all $\nu \in[a, r]$ and $\ell^{\prime} \in L, \mathrm{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}, r}\left(\ell^{\prime}, \nu\right)=\mathrm{Va}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\prime}, \nu\right)$. Furthermore, fake-optimal $N C$-strategies and optimal $F P$-strategies in $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$ are also fake-optimal and optimal over $[a, r]$ in $\mathcal{G}_{L^{\prime}, r}$.

Before proving this result, we start with an auxiliary lemma showing a property of the rates of change of the value functions associated to non-urgent locations

Lemma 14. Let $\mathcal{G}$ be an $r-S P T G, \ell$ and $\ell^{\prime}$ be non-urgent locations of Min and Max, respectively. Then for all $0 \leqslant \nu<\nu^{\prime} \leqslant r$ :

$$
\operatorname{slope}_{\mathcal{G}}^{\ell}\left(\nu, \nu^{\prime}\right) \geqslant-\pi(\ell) \quad \text { and } \quad \operatorname{slope}_{\mathcal{G}}^{\ell}\left(\nu, \nu^{\prime}\right) \leqslant-\pi\left(\ell^{\prime}\right)
$$

Proof. For the location $\ell$, the inequality rewrites in

$$
\operatorname{Val}_{\mathcal{G}}(\ell, \nu) \leqslant\left(\nu^{\prime}-\nu\right) \pi(\ell)+\operatorname{Va}_{\mathcal{G}}\left(\ell, \nu^{\prime}\right) .
$$

Using the upper definition of the value (thanks to the determinacy result of Theorem 11, it suffices to prove, for all $\varepsilon>0$, the existence of a strategy $\sigma_{\text {Min }}$ such that for all strategies $\sigma_{\text {Max }}$ of the opponent

$$
\operatorname{Price}\left(\operatorname{Play}\left((\ell, \nu), \sigma_{\operatorname{Min}}, \sigma_{\operatorname{Max}}\right)\right) \leqslant\left(\nu^{\prime}-\nu\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, \nu^{\prime}\right)+\varepsilon
$$

First, the definition of the value implies the existence of a strategy $\sigma_{\text {Min }}^{\prime}$ such that for all strategies $\sigma_{\text {Max }}$

$$
\operatorname{Price}\left(\operatorname{Play}\left(\left(\ell, \nu^{\prime}\right), \sigma_{\mathrm{Min}}^{\prime}, \sigma_{\mathrm{Max}}\right)\right) \leqslant \operatorname{Val}_{\mathcal{G}}\left(\ell, \nu^{\prime}\right)+\varepsilon
$$

Then, $\sigma_{\text {Min }}$ can be obtained by playing from $(\ell, \nu)$, at the first turn, as prescribed by $\sigma_{\text {Min }}^{\prime}$ but delaying $\nu^{\prime}-\nu$ time units more (that we are allowed to do since $\ell$ is non-urgent), and, for other turns, directly like $\sigma_{\text {Min }}^{\prime}$. A similar reasoning allows us to obtain the result for $\ell^{\prime}$.

Now, we show that, even if the locations in $L^{\prime}$ are turned into urgent locations, we may still obtain for them a similar result of the rates of change as the one of Lemma 14.

Lemma 15. For all locations $\ell \in L^{\prime} \cap L_{\text {Min }}$ (respectively, $\ell \in L^{\prime} \cap L_{\text {Max }}$ ), and $\nu \in[0, r], \mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell, \nu) \leqslant(r-\nu) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}(\ell, r)\left(\right.$ respectively, $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell, \nu) \geqslant$ $\left.(r-\nu) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}(\ell, r)\right)$.
Proof. It suffices to notice that from ( $\ell, \nu$ ), Min (respectively, Max) may choose
to go directly in $\ell^{f}$ ensuring the value $(r-\nu) \pi(\ell)+\mathrm{Val}_{\mathcal{G}}(\ell, r)$.

Proof of Proposition 13. Let $\sigma_{\mathrm{Min}}$ and $\sigma_{\mathrm{Max}}$ be a fake-optimal NC-strategy of Min and an optimal FP-strategy of Max in $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$, respectively. Notice that both strategies are also well-defined finite positional strategies in $\mathcal{G}_{L^{\prime}, r}$.

First, let us show that $\sigma_{\text {Min }}$ is indeed an NC-strategy in $\mathcal{G}_{L^{\prime}, r}$. Take a finite play $\left(\ell_{0}, \nu_{0}\right) \xrightarrow{c_{0}} \cdots \xrightarrow{c_{k-1}}\left(\ell_{k}, \nu_{k}\right)$, of length $k \geqslant 2$, that conforms with $\sigma_{\text {Min }}$ in $\mathcal{G}_{L^{\prime}, r}$, and with $\ell_{0}=\ell_{k}$ and $\nu_{0}, \nu_{k}$ in the same interval $I \operatorname{of} \operatorname{int}\left(\sigma_{\mathrm{Min}}\right)$. For every $\ell_{i}$ that is in $L_{\text {Min }}$, and $\nu \in I, \sigma_{\text {Min }}\left(\ell_{i}, \nu\right)$ must have a 0 delay, otherwise $\nu_{k}$ would not be in the same interval as $\nu_{0}$. Thus, the play $\left(\ell_{0}, \nu_{0}\right) \xrightarrow{c_{0}^{\prime}} \cdots \xrightarrow{c_{k-1}^{\prime}}\left(\ell_{k}, \nu_{0}\right)$ also conforms with $\sigma_{\text {Min }}$ (with possibly different prices). Furthermore, as all the delays are 0 we are sure that this play is also a valid play in $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$, in which $\sigma_{\text {Min }}$ is an NC-strategy. Therefore, $\pi\left(\ell_{0}, \ell_{1}\right)+\cdots+\pi\left(\ell_{k-1}, \ell_{k}\right) \leqslant-1$, and $\sigma_{\text {Min }}$ is an NC-strategy in $\mathcal{G}_{L^{\prime}, r}$.

We now show the result for $\ell \in L_{\text {Min }}$. The proof for $\ell \in L_{\text {Max }}$ is a straightforward adaptation. Notice that every play in $\mathcal{G}_{L^{\prime}, r}$ that conforms with $\sigma_{\text {Min }}$ is also a play in $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$ that conforms with $\sigma_{\text {Min }}$, as $\sigma_{\text {Min }}$ is defined in $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$ and thus plays with no delay in location $\ell$. Thus, for all $\nu \in[a, r]$ and $\ell^{\prime} \in L$, by the optimality result of Lemma 4 ,

$$
\begin{equation*}
\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\prime}, \nu\right) \leqslant \operatorname{fake}_{\mathcal{G}_{L^{\prime}, r}}^{\sigma_{\text {Min }}}\left(\ell^{\prime}, \nu\right)=\operatorname{fake}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}^{\sigma_{\text {Min }}}\left(\ell^{\prime}, \nu\right)=\operatorname{Val}_{\mathcal{G}_{\left.L^{\prime} \cup\{ \}\right\}, r}}\left(\ell^{\prime}, \nu\right) . \tag{4}
\end{equation*}
$$

To obtain that $\mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\prime}, \nu\right)=\mathrm{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right)$, it remains to show the reverse inequality. To that extent, let $\rho$ be a finite play in $\mathcal{G}_{L^{\prime}, r}$ that conforms with $\sigma_{\mathrm{Max}}$, starts in a configuration $\left(\ell^{\prime}, \nu\right)$ with $\nu \in[a, r]$, and ends in a final location. We show by induction on the length of $\rho$ that $\operatorname{Price}(\rho) \geqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right)$. If $\rho$ has size 1 then $\ell^{\prime}$ is a final configuration and $\operatorname{Price}(\rho)=\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right)=$ $\varphi_{\ell^{\prime}}^{\prime}(\nu)$.

Otherwise $\rho=\left(\ell^{\prime}, \nu\right) \xrightarrow{c} \rho^{\prime}$ where $\rho^{\prime}$ is a run that conforms with $\sigma_{\text {Max }}$, starting in a configuration ( $\ell^{\prime \prime}, \nu^{\prime \prime}$ ) and ending in a final configuration. By induction hypothesis, we have $\operatorname{Price}\left(\rho^{\prime}\right) \geqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right)$. We now distinguish three cases, the two first being immediate:

- If $\ell^{\prime} \in L_{\mathrm{Max}}$, then $\sigma_{\mathrm{Max}}\left(\ell^{\prime}, \nu\right)$ leads to the next configuration $\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right)$, thus

$$
\begin{aligned}
\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right) & =\operatorname{Price}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\left(\ell^{\prime}, \nu\right), \sigma_{\mathrm{Max}}\right) \\
& =c+\operatorname{Price}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right), \sigma_{\mathrm{Max}}\right) \\
& \leqslant c+\operatorname{Price}\left(\rho^{\prime}\right)=\operatorname{Price}(\rho) .
\end{aligned}
$$

- If $\ell^{\prime} \in L_{\text {Min }}$, and $\ell^{\prime} \neq \ell$ or $\nu^{\prime \prime}=\nu$, we have that $\left(\ell^{\prime}, \nu\right) \xrightarrow{c}\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right)$ is a valid transition in $\mathcal{G}^{\prime}$. Therefore, $\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right) \leqslant c+\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right)$, hence

$$
\operatorname{Price}(\rho)=c+\operatorname{Price}\left(\rho^{\prime}\right) \geqslant c+\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right) \geqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right)
$$

- Finally, if $\ell^{\prime}=\ell$ and $\nu^{\prime \prime}>\nu$, then $c=\left(\nu^{\prime \prime}-\nu\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime \prime}\right)$. As $\left(\ell, \nu^{\prime \prime}\right) \xrightarrow{\pi\left(\ell, \ell^{\prime \prime}\right)}$ $\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right)$ is a valid transition in $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$, we have $\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell, \nu^{\prime \prime}\right) \leqslant$ $\pi\left(\ell, \ell^{\prime \prime}\right)+\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right)$. Furthermore, since $\nu^{\prime \prime} \in[a, r]$, we can use (3) to obtain

$$
\begin{aligned}
\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}(\ell, \nu) & \leqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell, \nu^{\prime \prime}\right)+\left(\nu^{\prime \prime}-\nu\right) \pi(\ell) \\
& \leqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right)+\pi\left(\ell, \ell^{\prime \prime}\right)+\left(\nu^{\prime \prime}-\nu\right) \pi(\ell) .
\end{aligned}
$$

Therefore

$$
\begin{aligned}
\operatorname{Price}(\rho) & =\left(\nu^{\prime \prime}-\nu\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime \prime}\right)+\operatorname{Price}\left(\rho^{\prime}\right) \\
& \geqslant\left(\nu^{\prime \prime}-\nu\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime \prime}\right)+\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime \prime}, \nu^{\prime \prime}\right) \geqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right) .
\end{aligned}
$$

This concludes the induction. As a consequence,

$$
\inf _{\sigma_{\text {Min }}^{\prime} \in \operatorname{Strat}_{\text {Min }}\left(\mathcal{G}_{L^{\prime}, r}\right)} \operatorname{Price}_{\mathcal{G}_{L^{\prime}, r}^{\prime}}\left(\operatorname{Play}\left(\left(\ell^{\prime}, \nu\right), \sigma_{\text {Min }}^{\prime}, \sigma_{\text {Max }}\right)\right) \geqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right)
$$

for all locations $\ell^{\prime}$ and $\nu \in[a, r]$, which finally proves that $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\prime}, \nu\right) \geqslant$ $\operatorname{Val}_{\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}}\left(\ell^{\prime}, \nu\right)$. Fake-optimality of $\sigma_{\text {Min }}$ over $[a, r]$ in $\mathcal{G}_{L^{\prime} \cup\{\ell\}, r}$ is then obtained by (4).

Given an SPTG $\mathcal{G}$ and some finitely optimal $\mathcal{G}_{L^{\prime}, r}$, we now characterise precisely the left endpoint of the maximal interval ending in $r$ where the value functions of $\mathcal{G}$ and $\mathcal{G}_{L^{\prime}, r}$ coincide, with the operator $\operatorname{left}_{L^{\prime}}:(0,1] \rightarrow[0,1]$ (or simply left, if $L^{\prime}$ is clear) defined as:

$$
\operatorname{left}_{L^{\prime}}(r)=\inf \left\{r^{\prime} \leqslant r \mid \forall \ell \in L \forall \nu \in\left[r^{\prime}, r\right] \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)\right\} .
$$

By continuity of the value (Theorem2 2 , this infimum exists and $\mathrm{Val}_{\mathcal{G}}\left(\ell, \operatorname{left}_{L^{\prime}}(r)\right)=$ $\operatorname{Val}_{\mathcal{G}_{L^{\prime} r} r}\left(\ell, \operatorname{left}_{L^{\prime}}(r)\right)$. Moreover, $\mathrm{Val}_{\mathcal{G}}(\ell)$ is a cost function on $[\operatorname{left}(r), r]$, since $\mathcal{G}_{L^{\prime}, r}$ is finitely optimal. However, this definition of left $(r)$ is semantical. Yet, building on the ideas of Proposition 13 , we can effectively compute left $(r)$, given $\mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}$. We claim that left lit $_{L^{\prime}}(r)$ is the minimal valuation such that for all locations $\ell \in L^{\prime} \cap L_{\text {Min }}$ (respectively, $\ell \in L^{\prime} \cap L_{\text {Max }}$ ), the slopes of the affine sections of the cost function $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell)$ on $[\operatorname{left}(r), r]$ are at least (at most) $-\pi(\ell)$. Hence, left $(r)$ can be obtained (see Figure 7) by inspecting iteratively, for all $\ell$ of Min (respectively, Max), the slopes of $\mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell)$ by decreasing valuations until we find a piece with a slope greater than $-\pi(\ell)$ (respectively, smaller than $-\pi(\ell)$ ). This enumeration of the slopes is effective as $\mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}$ has finitely many pieces, by hypothesis. Moreover, this guarantees that left $(r)<r$, as shown in the following lemma.

Lemma 16. Let $\mathcal{G}$ be an $S P T G, L^{\prime} \subseteq L \backslash L_{u}$, and $r \in(0,1]$, such that $\mathcal{G}_{L^{\prime \prime}, r}$ is finitely optimal for all $L^{\prime \prime} \subseteq L^{\prime}$. Then, left $L_{L^{\prime}}(r)$ is the minimal valuation such that for all locations $\ell \in L^{\prime} \cap L_{\text {Min }}$ (respectively, $\ell \in L^{\prime} \cap L_{\text {Max }}$ ), the slopes of the affine sections of the cost function $\mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell)$ on $[l e f t(r), r]$ are at least (respectively, at most) $-\pi(\ell)$. Moreover, left $(r)<r$.

Proof. Since $\mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell)=\mathrm{Va}_{\mathcal{G}}(\ell)$ on $[\operatorname{left}(r), r]$, and as $\ell$ is non-urgent in $\mathcal{G}$, Lemma 14 states that all the slopes of $\operatorname{Val}_{\mathcal{G}}(\ell)$ are at least (respectively, at most) $-\pi(\ell)$ on $[\operatorname{left}(r), r]$.

We now show the minimality property by contradiction. Therefore, let $r^{\prime}<$ left $(r)$ such that all cost functions $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell)$ are affine on $\left[r^{\prime}\right.$, left $\left.(r)\right]$, and assume that for all $\ell \in L^{\prime} \cap L_{\text {Min }}$ (respectively, $\ell \in L^{\prime} \cap L_{\mathrm{Max}}$ ), the slopes of $\mathrm{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell)$ on $\left[r^{\prime}, \operatorname{left}(r)\right]$ are at least (respectively, at most) $-\pi(\ell)$. Hence, this property holds on $\left[r^{\prime}, r\right]$. Then, by applying Proposition $\sqrt[13]{ }\left|L^{\prime}\right|$ times (here, we use the finite optimality of the games $\mathcal{G}_{L^{\prime \prime}, r}$ with $L^{\prime \prime} \subseteq L^{\prime}$ ), we have that for all $\nu \in\left[r^{\prime}, r\right]$ $\operatorname{Val}_{\mathcal{G}_{r}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}(\ell, \nu)$. Using Lemma 12 , we also know that for all $\nu \leqslant r$,


Figure 7: In this example $L^{\prime}=\left\{\ell^{\star}\right\}$ and $\ell^{\star} \in L_{\text {Min }}$. left $(r)$ is the leftmost point such that all slopes on its right are smaller than or equal to $-\pi\left(\ell^{\star}\right)$ in the graph of $\mathrm{Va}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}, \nu\right)$. Dashed lines have slope $-\pi\left(\ell^{\star}\right)$.
and $\ell, \operatorname{Val}_{\mathcal{G}_{r}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)$. Thus, $\operatorname{Val}_{\mathcal{G}_{r, L^{\prime}}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)$. As $r^{\prime}<\operatorname{left}(r)$, this contradicts the definition of left $L_{L^{\prime}}(r)$.

We finally prove that $\operatorname{left}(r)<r$. This is immediate in case $\operatorname{left}(r)=0$, since $r>0$. Otherwise, from the result obtained previously, we know that there exists $r^{\prime}<\operatorname{left}(r)$, and $\ell^{\star} \in L^{\prime}$ such that $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}\right)$ is affine on $\left[r^{\prime}\right.$, left $(r)$ ] of slope smaller (respectively, greater) than $-\pi\left(\ell^{\star}\right)$ if $\ell^{\star} \in L_{\text {Min }}$ (respectively, $\ell^{\star} \in L_{\text {Max }}$ ), i.e.

$$
\begin{cases}\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r^{\prime}\right)>\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, \operatorname{left}(r)\right)+\left(\operatorname{left}(r)-r^{\prime}\right) \pi\left(\ell^{\star}\right) & \text { if } \ell^{\star} \in L_{\text {Min }} \\ \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r^{\prime}\right)<\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, \operatorname{left}(r)\right)+\left(\operatorname{left}(r)-r^{\prime}\right) \pi\left(\ell^{\star}\right) & \text { if } \ell^{\star} \in L_{\mathrm{Max}} .\end{cases}
$$

From Lemma 15 we also know that

$$
\begin{cases}\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r^{\prime}\right) \leqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r\right)+\left(r-r^{\prime}\right) \pi\left(\ell^{\star}\right) & \text { if } \ell^{\star} \in L_{\mathrm{Min}} \\ \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r^{\prime}\right) \geqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r\right)+\left(r-r^{\prime}\right) \pi\left(\ell^{\star}\right) & \text { if } \ell^{\star} \in L_{\mathrm{Max}}\end{cases}
$$

Both equations combined imply

$$
\begin{cases}\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r\right)>\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, \operatorname{left}(r)\right)+(\operatorname{left}(r)-r) \pi\left(\ell^{\star}\right) & \text { if } \ell^{\star} \in L_{\operatorname{Min}} \\ \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, r\right)<\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r}}\left(\ell^{\star}, \operatorname{left}(r)\right)+(\operatorname{left}(r)-r) \pi\left(\ell^{\star}\right) & \text { if } \ell^{\star} \in L_{\operatorname{Max}}\end{cases}
$$

which is not possible if left $(r)=r$.
Thus, one can reconstruct $\mathrm{Val}_{\mathcal{G}}$ on $\left[\inf _{i} r_{i}, r_{0}\right]$ from the value functions of the (potentially infinite) sequence of games $\mathcal{G}_{L^{\prime}, r_{0}}, \mathcal{G}_{L^{\prime}, r_{1}}, \ldots$ where $r_{i+1}=\operatorname{left}\left(r_{i}\right)$ for all $i$ such that $r_{i}>0$, for all possible choices of non-urgent locations $L^{\prime}$. Next, we will define two different ways of choosing $L^{\prime}$ : the former to prove finite optimality of all SPTGs, the latter to obtain an algorithm to solve them.

### 5.2. SPTGs are finitely optimal.

To prove finite optimality of all SPTGs we reason by induction on the number of non-urgent locations and instantiate the previous results to the case where

Assume by contradiction that there exists an index $i$ such that $\nu<\nu_{i}$ and let $i$ be the smallest of such indices. For each $j<i$, if $\ell_{j} \in L_{\text {Min }}$, let $(t, \delta)=$ $\sigma_{\text {Min }}^{\prime}\left(\ell_{j}, \nu\right)$ and $\left(t^{\prime}, \delta^{\prime}\right)=\sigma_{\text {Min }}^{\prime}\left(\ell_{j}, \nu^{\prime}\right)$. Similarly, if $\ell_{j} \in L_{\text {Max }}$, we let $(t, \delta)=$ $\sigma_{\text {Max }}\left(\ell_{j}, \nu\right)$ and $\left(t^{\prime}, \delta^{\prime}\right)=\sigma_{\text {Max }}\left(\ell_{j}, \nu^{\prime}\right)$. As $I^{\prime}$ is contained in an interval of $\sigma_{\text {Min }}^{\prime}$ and $\sigma_{\text {Max }}$, we have $\delta=\delta^{\prime}$ and either $t=t^{\prime}=0$, or $\nu+t=\nu^{\prime}+t^{\prime}$. Applying this result for all $j<i$, we obtain that $\left(\ell_{0}, \nu^{\prime}\right) \xrightarrow{c_{0}^{\prime}} \cdots\left(\ell_{i-1}, \nu^{\prime}\right) \xrightarrow{c_{i-1}^{\prime}}\left(\ell_{i}, \nu_{i}\right) \xrightarrow{c_{i}} \cdots\left(\ell_{k}, \nu_{k}\right)$ is a prefix of $\operatorname{Play}\left(\left(\ell^{\star}, \nu^{\prime}\right), \sigma_{\text {Min }}^{\prime}, \sigma_{\mathrm{Max}}\right)$ : notice moreover that, as before, this prefix has price $\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}, \nu^{\prime}\right)$. In particular,
$\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}, \nu^{\prime}\right)=\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}, \nu\right)-\left(\nu^{\prime}-\nu\right) \pi\left(\ell_{i-1}\right) \leqslant \operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}, \nu\right)-\left(\nu^{\prime}-\nu\right) \pi\left(\ell^{\star}\right)$
which implies that the slope of $\mathrm{Val}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}\right)$ is at most $-\pi\left(\ell^{\star}\right)$, and therefore contradicts the hypothesis. As a consequence, we have that $\nu_{i}=\nu$ for all $i$.

Again by contradiction, assume now that $\ell_{k}=\ell^{f}$ for some $\ell \in L \backslash L_{u}$. By the same reasoning as before, we then would have $\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}, \nu^{\prime}\right)=\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r}}\left(\ell^{\star}, \nu\right)-$ $\left(\nu^{\prime}-\nu\right) \pi(\ell)$, which again contradicts the hypothesis.

Therefore, $\ell_{k} \in L_{f}$. Suppose, for a contradiction, that the prefix $\left(\ell_{0}, \nu\right) \xrightarrow{c_{0}}$ $\cdots\left(\ell_{k}, \nu\right)$ contains a cycle. Since $\sigma_{\text {Min }}^{\prime}$ is a switching strategy and $\sigma_{\text {Max }}$ is a memoryless strategy, this implies that the cycle is contained in the part of $\sigma_{\text {Min }}^{\prime}$ where the decision is taken by the strategy $\sigma_{\text {Min }}$ : since it is an NC-strategy,

[^2] $\left[r^{\prime}, r_{i+1}\right]$ and its slope is smaller that $-\pi\left(\ell^{\star}\right)$, i.e. $\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i}}}\left(\ell^{\star}, r_{i+1}\right)<\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i}}}\left(r^{\prime}\right)-$ $\left(r_{i+1}-r^{\prime}\right) \pi\left(\ell^{\star}\right)$. Lemma 15 also ensures that $\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i}}}\left(\ell^{\star}, r^{\prime}\right) \leqslant \operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{i}\right)+$ $\left(r_{i}-r^{\prime}\right) \pi\left(\ell^{\star}\right)$. Combining both inequalities allows us to conclude.

We now turn to the proof of the second item, showing the stationarity of sequence $\left(r_{i}\right)_{i \geqslant 0}$. We consider first the case where $\ell^{\star} \in L_{\mathrm{Max}}$. Let $i>0$ such that $r_{i} \neq 0$ (if there exist no such $i$ then $r_{1}=0$ ). Recall from Lemma 16 that there exists $r_{i}^{\prime}<r_{i}$ such that $\mathrm{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}\right)$ is affine on $\left[r_{i}^{\prime}, r_{i}\right]$, of slope greater than $-\pi\left(\ell^{\star}\right)$. In particular,

$$
\frac{\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}, r_{i}\right)-\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}, r_{i}^{\prime}\right)}{r_{i}-r_{i}^{\prime}}>-\pi\left(\ell^{\star}\right) .
$$

Lemma 17 states that on $\left[r_{i}^{\prime}, r_{i}\right]$, $\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}\right)$ is equal to some $f_{i} \in \mathrm{~F}_{\mathcal{G}}$. As $f_{i}$ is an affine function, $f_{i}\left(r_{i}\right)=\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}, r_{i}\right)$, and $f_{i}\left(r_{i}^{\prime}\right)=\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}, r_{i}^{\prime}\right)$. Thus, for all $\nu$,

$$
f_{i}(\nu)=\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}, r_{i}\right)+\frac{\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i}-1}}\left(\ell^{\star}, r_{i}^{\prime}\right)-\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}, r_{i}\right)}{r_{i}-r_{i}^{\prime}}\left(r_{i}-\nu\right) .
$$

Since $\mathcal{G}_{\ell^{\star}, r_{i-1}}$ is assumed to be finitely optimal, we know that $\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}\left(\ell^{\star}, r_{i}\right)=$ $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{i}\right)$, by definition of $r_{i}=\operatorname{left}_{\ell^{\star}}\left(r_{i-1}\right)$. Therefore, combining both equalities above, for all valuations $\nu<r_{i}$, we have $f_{i}(\nu)<\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{i}\right)+\pi\left(\ell^{\star}\right)\left(r_{i}-\nu\right)$.


Figure 8: The case $\ell^{\star} \in L_{\mathrm{Max}}$ : a geometric proof of $f_{i} \neq f_{j}$. The dotted lines represents $f_{i}$ and $f_{j}$, the dashed lines have slope $-\pi\left(\ell^{\star}\right)$, and the plain line depicts $\mathrm{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$. Because the slope of $f_{i}$ is strictly smaller than $-\pi\left(\ell^{\star}\right)$, and the value at $r_{j}$ is above the dashed line it cannot be the case that $f_{i}\left(r_{j}\right)=\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{j}\right)=f_{j}\left(r_{j}\right)$.

Consider then $j>i$ such that $r_{j} \neq 0$. We claim that $f_{j} \neq f_{i}$. Indeed, we have $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{j}\right)=f_{j}\left(r_{j}\right)$. As, in $\mathcal{G}, \ell^{\star}$ is a non-urgent location, Lemma 14 ensures that

$$
\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{j}\right) \geqslant \operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{i}\right)+\pi\left(\ell^{\star}\right)\left(r_{i}-r_{j}\right) .
$$

As for all $i^{\prime}, \operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{i^{\prime}}\right)=f_{i^{\prime}}\left(r_{i^{\prime}}\right)$, the equality above is equivalent to $f_{j}\left(r_{j}\right) \geqslant$ $f_{i}\left(r_{i}\right)+\pi\left(\ell^{\star}\right)\left(r_{i}-r_{j}\right)$. Recall that $f_{i}$ has a slope strictly greater that $-\pi\left(\ell^{\star}\right)$, therefore $f_{i}\left(r_{j}\right)<f_{i}\left(r_{i}\right)+\pi\left(\ell^{\star}\right)\left(r_{i}-r_{j}\right) \leqslant f_{j}\left(r_{j}\right)$. As a consequence $f_{i} \neq f_{j}$ (this is depicted in Figure 8).

Therefore, there cannot be more than $\left|\mathrm{F}_{\mathcal{G}}\right|+1$ non-null elements in the sequence $r_{0} \geqslant r_{1} \geqslant \cdots$, which proves that there exists $i \leqslant\left|\mathrm{~F}_{\mathcal{G}}\right|+2$ such that $r_{i}=0$.

We continue with the case where $\ell^{\star} \in L_{\text {Min }}$. Let $r_{\infty}=\inf \left\{r_{i} \mid i \geqslant 0\right\}$. In this case, we look at the affine parts of $\overline{\mathrm{Val}_{\mathcal{G}}\left(\ell^{\star}\right)}$ with a slope greater than $-\pi\left(\ell^{\star}\right)$, and we show that there can only be finitely many such segments in $\left[r_{\infty}, 1\right]$. We then show that there is at least one such segment contained in $\left[r_{i+1}, r_{i}\right]$ for all $i$, bounding the size of the sequence.

In the following, we call segment every interval $[a, b] \subset\left(r_{\infty}, 1\right]$ such that $a$ and $b$, are two consecutive cutpoints of the cost function $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$ over $\left(r_{\infty}, 1\right]$. Recall that it means that $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$ is affine on $[a, b]$, and if we let $a^{\prime}$ be the greatest cutpoint smaller than $a$, and $b^{\prime}$ the smallest cutpoint greater than $b$, the slopes of $\mathrm{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$ on $\left[a^{\prime}, a\right]$ and $\left[b, b^{\prime}\right]$ are different from the slope on $[a, b]$. We abuse the notations by referring to the slope of a segment $[a, b]$ for the slope of $\mathrm{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$ on $[a, b]$ and simply call cutpoint a cutpoint of $\mathrm{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$.

To every segment $[a, b]$ with a slope greater than $-\pi\left(\ell^{\star}\right)$, we associate a function $f_{[a, b]} \in \mathrm{F}_{\mathcal{G}}$ as follows. Let $i$ be the smallest index such that $[a, b] \cap$ $\left[r_{i+1}, r_{i}\right]$ is a non singleton interval $\left[a^{\prime}, b^{\prime}\right]$. Lemma 17 ensures that there exists $f_{[a, b]} \in \mathrm{F}_{\mathcal{G}}$ such that for all $\nu \in\left[a^{\prime}, b^{\prime}\right], \operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, \nu\right)=f_{[a, b]}(\nu)$.

Consider now two disjoint segments $[a, b]$ and $[c, d]$ with a slope strictly greater than $-\pi\left(\ell^{\star}\right)$, and assume that $f_{[a, b]}=f_{[c, d]}$ (in particular both segments


Figure 9: In order for the segments $[a, b]$ and $[c, d]$ to be aligned, there must exist a segment with a biggest slope crossing $f_{[a, b]}$ (represented by a dashed line) between $b$ and $c$.
have the same slope). Without loss of generality, assume that $b<c$. We claim that there exists a segment $[e, g]$ in-between $[a, b]$ and $[c, d]$ with a slope greater than the slope of $[c, d]$, and that $f_{[e, g]}$ and $f_{[a, b]}$ intersect over $[b, c]$, in a point of abscisse $x$, i.e. $x \in[b, c]$ verifies $f_{[e, g]}(x)=f_{[a, b]}(x)$ (depicted in Figure 9]. We prove it now.

Let $\alpha$ be the greatest cutpoint smaller than $c$. We know that the slope of $[\alpha, c]$ is different from the one of $[c, d]$. If it is greater then define $e=\alpha$ and $x=g=c$, those indeed satisfy the property. If the slope of $[\alpha, c]$ is smaller than the one of $[c, d]$, then for all $\nu \in[\alpha, c), \operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, \nu\right)>f_{[c, d]}(\nu)$. Let $x$ be the greatest point in $[b, \alpha]$ such that $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, x\right)=f_{[c, d]}(x)$. We know that it exists since $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, b\right)=f_{[c, d]}(b)$, and $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$ is continuous. Observe that $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, \nu\right)>$ $f_{[c, d]}(\nu)$, for all $x<\nu<c$. Finally, let $g$ be the smallest cutpoint of $\mathrm{Va}_{\mathcal{G}}\left(\ell^{\star}\right)$ strictly greater than $x$, and $e$ the greatest cutpoint of $\mathrm{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$ smaller than or equal to $x$. By construction $[e, g]$ is a segment that contains $x$. The slope of the segment $[e, g]$ is $s_{[e, g]}=\frac{\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, g\right)-\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, x\right)}{g-x}$, and the slope of the segment $[c, d]$ is equal to $s_{[c, d]}=\frac{f_{[c, d]}(g)-f_{[c, d]}(x)}{g-x}$. Remembering that $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, x\right)=f_{[c, d]}(x)$, and that $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, g\right)>f_{[c, d]}(g)$ since $g \in(x, c)$, we obtain that $s_{[e, g]}>s_{[c, d]}$. Finally, since $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, x\right)=f_{[c, d]}(x)=f_{[e, g]}(x)$, it is indeed the abscisse of the intersection point of $f_{[c, d]}=f_{[a, b]}$ and $f_{[e, g]}$, which concludes the proof of the previous claim.

For every function $f \in \mathrm{~F}_{\mathcal{G}}$, there are less than $\left|\mathrm{F}_{\mathcal{G}}\right|$ intersection points between $f$ and the other functions of $\mathbf{F}_{\mathcal{G}}$ (at most one for each pair $\left(f, f^{\prime}\right)$ ). If $f$ has a slope greater than $-\pi\left(\ell^{\star}\right)$, thanks to the previous paragraph, we know that there are at most $\left|\mathrm{F}_{\mathcal{G}}\right|$ segments $[a, b]$ such that $f_{[a, b]}=f$. Summing over all possible functions $f$, there are at most $\left|F_{\mathcal{G}}\right|^{2}$ segments with a slope greater than $-\pi\left(\ell^{\star}\right)$.

Now, we link those segments with the valuations $r_{i}$ 's, for $i>0$. By item


Figure 10: The case $\ell^{\star} \in L_{\mathrm{Min}}$ : as the value at $r_{i+1}$ is strictly below $\mathrm{Va}_{\mathcal{G}}\left(r_{i}\right)+\pi\left(\ell^{\star}\right)\left(r_{i}-r_{i+1}\right)$, as the slope on the left of $r_{i}$ and of $r_{i+1}$ is $-\pi\left(\ell^{\star}\right)$, there must exist a segment (represented with a double line) with slope greater than $-\pi\left(\ell^{\star}\right)$ in $\left[r_{i+1}, r_{i}\right)$.
(i), thanks to the finite-optimality of $\mathcal{G}_{\ell^{\star}, r_{i}}, \operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{i+1}\right)<\left(r_{i}-r_{i+1}\right) \pi\left(\ell^{\star}\right)+$ $\operatorname{Val}_{\mathcal{G}}\left(\ell^{\star}, r_{i}\right)$. Furthermore, Proposition 20 states that the slope of the segment directly on the left of $r_{i}$ is equal to $-\pi\left(\ell^{\star}\right)$. With the previous inequality in mind, this cannot be the case if $\mathrm{Val}_{\mathcal{G}}\left(\ell^{\star}\right)$ is affine over the whole interval $\left[r_{i+1}, r_{i}\right]$. Thus, there exists a segment $[a, b]$ of slope strictly greater than $-\pi\left(\ell^{\star}\right)$ such that $b \in\left[r_{i+1}, r_{i}\right]$. As we also know that the slope left to $r_{i+1}$ is $-\pi\left(\ell^{\star}\right)$, it must be the case that $a \in\left[r_{i+1}, r_{i}\right]$. Hence, we have shown that in-between $r_{i+1}$ and $r_{i}$, there is always a segment (this is depicted in Figure 10). As the number of such segments is bounded by $\left|\mathrm{F}_{\mathcal{G}}\right|^{2}$, we know that the sequence $r_{i}$ is stationary in at most $\left|\mathrm{F}_{\mathcal{G}}\right|^{2}+1$ steps, i.e. that there exists $i \leqslant\left|\mathrm{~F}_{\mathcal{G}}\right|^{2}+1$ such that $r_{i}=0$.

By iterating this construction, we make all locations urgent iteratively, and obtain:

Theorem 19. Every $S P T G \mathcal{G}$ is finitely optimal and for all locations $\ell, \operatorname{Val}_{\mathcal{G}}(\ell)$ has at most $O\left(\left(\Pi^{\mathrm{tr}}|L|^{2}\right)^{2|L|+2}\right)$ cutpoints.

Proof. As announced, we show by induction on $n \geqslant 0$ that every $r$-SPTG $\mathcal{G}$ with $n$ non-urgent locations is finitely optimal, and that the number of cutpoints of $\operatorname{Val}_{\mathcal{G}}(\ell)$ is at most $O\left(\left(\Pi^{\operatorname{tr}}\left(\left|L_{f}\right|+n^{2}\right)\right)^{2 n+2}\right)$, which suffices to show the above bound, since $\left|L_{f}\right|+n^{2} \leqslant|L|^{2}$.

The base case $n=0$ is given by Proposition 11. Now, assume that $\mathcal{G}$ has at least one non-urgent location, and consider $\ell^{\star}$ one with minimum pricerate. By induction hypothesis, all $r^{\prime}$-SPTGs $\mathcal{G}_{\ell^{\star}, r^{\prime}}$ are finitely optimal for all $r^{\prime} \in[0, r]$. Let $r_{0}>r_{1}>\cdots$ be the decreasing sequence defined by $r_{0}=r$ and $r_{i}=\operatorname{left}_{\ell^{\star}}\left(r_{i-1}\right)$ for all $i \geqslant 1$. By Lemma 18 there exists $j \leqslant\left|\mathrm{~F}_{\mathcal{G}}\right|^{2}+2$ such that $r_{j}=0$. Moreover, for all $0<i \leqslant j, \mathrm{Val}_{\mathcal{G}}=\operatorname{Val}_{\mathcal{G}_{\ell^{\star}, r_{i-1}}}$ on $\left[r_{i}, r_{i-1}\right]$ by definition of $r_{i}=\operatorname{left}_{\ell^{\star}}\left(r_{i-1}\right)$, so that $\mathrm{Val}_{\mathcal{G}}(\ell)$ is a cost function on this interval, for all $\ell$, and the number of cutpoints on this interval is bounded by $O\left(\left(\Pi^{\operatorname{tr}}\left(\left|L_{f}\right|+(n-1)^{2}+n\right)\right)^{2(n-1)+2}\right)=O\left(\left(\Pi^{\operatorname{tr}}\left(\left|L_{f}\right|+n^{2}\right)\right)^{2(n-1)+2}\right)$ by induction hypothesis (notice that maximal transition prices are the same in $\mathcal{G}$ and
$\mathcal{G}_{\ell^{\star}, r_{i-1}}$, but that we add $n$ more final locations in $\left.\mathcal{G}_{\ell^{\star}, r_{i-1}}\right)$. Adding the cutpoint 1 , summing over $i$ from 0 to $j \leqslant\left|\mathrm{~F}_{\mathcal{G}}\right|^{2}+2$, and observing that $\left|\mathrm{F}_{\mathcal{G}}\right| \leqslant 2 \Pi^{\mathrm{tr}}\left|L_{f}\right|$, we bound the number of cutpoints of $\operatorname{Val}_{\mathcal{G}}(\ell)$ by $O\left(\left(\Pi^{\operatorname{tr}}\left(\left|L_{f}\right|+n^{2}\right)\right)^{2 n+2}\right)$. Finally, we can reconstruct fake-optimal and optimal strategies in $\mathcal{G}$ from the fake-optimal and optimal strategies of $\mathcal{G}_{\ell^{\star}, r_{i}}$.

## 6. Algorithms to compute the value function

The finite optimality of SPTGs allows us to compute the value functions. The proof of Theorem 19 suggests a recursive algorithm to do so: from an SPTG $\mathcal{G}$ with minimal non-urgent location $\ell^{\star}$, solve recursively $\mathcal{G}_{\ell^{\star}, 1}, \mathcal{G}_{\ell^{\star}, \text { left }(1)}$, $\mathcal{G}_{\ell^{\star}, \operatorname{left}(\operatorname{left}(1))}$, etc. handling the base case where all locations are urgent with Algorithm 1. While our results above show that this is correct and terminates, we propose instead to solve - without the need for recursion - the sequence of games $\mathcal{G}_{L \backslash L_{u}, 1}, \mathcal{G}_{L \backslash L_{u}, \text { left(1) }}, \ldots$ i.e. making all locations urgent at once. Again, the arguments given above prove that this scheme is correct, but the key argument of Lemma 18 that ensures termination cannot be applied in this case. Instead, we rely on the following result, stating, that there will be at least one cutpoint of $\mathrm{Val}_{\mathcal{G}}$ in each interval $[\operatorname{left}(r), r]$. Observe that this lemma relies on the fact that $\mathcal{G}$ is finitely optimal, hence the need to first prove this fact independently with the sequence $\mathcal{G}_{\ell^{\star}, 1}, \mathcal{G}_{\ell^{\star}, \text { left }(1)}, \mathcal{G}_{\ell^{\star}, \text { left (left(1)) }}, \ldots$ Termination then follows from the fact that $\mathrm{Val}_{\mathcal{G}}$ has finitely many cutpoints by finite optimality.

Proposition 20. Let $r_{0} \in(0,1]$ such that $\mathcal{G}_{L^{\prime}, r_{0}}$ is finitely optimal. Suppose that $r_{1}=\operatorname{left}_{L^{\prime}}\left(r_{0}\right)>0$, and let $r_{2}=$ left $_{L^{\prime}}\left(r_{1}\right)$. There exists $r^{\prime} \in\left[r_{2}, r_{1}\right)$ and $\ell \in L^{\prime}$ such that
(i) $\mathrm{Val}_{\mathcal{G}}(\ell)$ is affine on $\left[r^{\prime}, r_{1}\right]$, of slope equal to $-\pi(\ell)$, and
(ii) $\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{1}\right) \neq \operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right)+\pi(\ell)\left(r_{0}-r_{1}\right)$.

As a consequence, $\operatorname{Val}_{\mathcal{G}}(\ell)$ has a cutpoint in $\left[r_{1}, r_{0}\right)$.
Proof. We denote by $r^{\prime}$ the smallest valuation (smaller than $r_{1}$ ) such that for all locations $\ell, \mathrm{Val}_{\mathcal{G}}(\ell)$ is affine over $\left[r^{\prime}, r_{1}\right]$. Then, the proof goes by contradiction: using Lemma 16, we assume that for all $\ell \in L^{\prime} \cap L_{\text {Min }}$ (respectively, $\ell \in L^{\prime} \cap L_{\text {Max }}$ )

- either $(\neg(i))$ the slope of $\operatorname{Val}_{\mathcal{G}}(\ell)$ on $\left[r^{\prime}, r_{1}\right]$ is greater (respectively, smaller) than $-\pi(\ell)$,
- or $((i) \wedge \neg(i i))$ for all $\nu \in\left[r^{\prime}, r_{1}\right], \operatorname{Val}_{\mathcal{G}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right)+\pi(\ell)\left(r_{0}-\nu\right)$.

Let $\sigma_{\text {Min }}^{0}$ and $\sigma_{\text {Max }}^{0}$ (respectively, $\sigma_{\text {Min }}^{1}$ and $\sigma_{\text {Max }}^{1}$ ) be a fake-optimal NCstrategy and an optimal FP-strategy in $\mathcal{G}_{L^{\prime}, r_{0}}$ (respectively, $\mathcal{G}_{L^{\prime}, r_{1}}$ ). Let $r^{\prime \prime}=$ $\max \left(\operatorname{pts}\left(\sigma_{\text {Min }}^{1}\right) \cup \operatorname{pts}\left(\sigma_{\text {Max }}^{1}\right)\right) \cap\left[r^{\prime}, r_{1}\right)$, so that strategies $\sigma_{\text {Min }}^{1}$ and $\sigma_{\text {Max }}^{1}$ have the same behaviour on all valuations of the interval ( $r^{\prime \prime}, r_{1}$ ), i.e. either always play urgently the same transition, or wait, in a non-urgent location, until reaching some valuation greater than or equal to $r_{1}$ and then play the same transition.

Observe preliminarily that for all $\ell \in L^{\prime} \cap L_{\operatorname{Min}}$ (respectively, $\ell \in L^{\prime} \cap L_{\mathrm{Max}}$ ), if on the interval $\left(r^{\prime \prime}, r_{1}\right), \sigma_{\text {Min }}^{1}$ (respectively, $\sigma_{\text {Max }}^{1}$ ) goes to $\ell^{f}$ then the slope on [ $r^{\prime \prime}, r_{1}$ ] (and thus on $\left[r^{\prime}, r_{1}\right]$ ) is $-\pi(\ell)$. Thus for such a location $\ell$, we know that (i) $\wedge \neg(i i)$ holds for $\ell$ (by letting $r^{\prime}$ be $r^{\prime \prime}$ ).

For other locations $\ell$, we will construct a new pair of NC- and FP-strategies $\sigma_{\text {Min }}$ and $\sigma_{\text {Max }}$ in $\mathcal{G}_{L^{\prime}, r_{0}}$ such that for all locations $\ell$ and valuations $\nu \in\left(r^{\prime \prime}, r_{1}\right)$

$$
\begin{equation*}
\operatorname{fake}_{\mathcal{G}_{L^{\prime}, r_{0}}}^{\sigma_{\text {Min }}}(\ell, \nu) \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu) \leqslant \operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left((\ell, \nu), \sigma_{\text {Max }}\right) . \tag{5}
\end{equation*}
$$

As a consequence, with Lemma 4 (over game $\mathcal{G}_{L^{\prime}, r_{0}}$ ), one would have that $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)$, which will raise a contradiction with the definition of $r_{1}$ as $\operatorname{left}_{L^{\prime}}\left(r_{0}\right)<r_{0}$ (by Lemma 16), and conclude the proof.

We only show the construction for $\sigma_{\text {Min }}$, as it is very similar for $\sigma_{\text {Max }}$. Strategy $\sigma_{\text {Min }}$ is obtained by combining strategies $\sigma_{\text {Min }}^{1}$ over $\left[0, r_{1}\right]$, and $\sigma_{\text {Min }}^{0}$ over $\left[r_{1}, r_{0}\right]$ : a special care has to be spent in case $\sigma_{\text {Min }}^{1}$ performs a jump to a location $\ell^{f}$, since then, in $\sigma_{\text {Min }}$, we rather glue this move with the decision of strategy $\sigma_{\mathrm{Min}}^{0}$ in $\left(\ell, r_{1}\right)$. Formally, let $(\ell, \nu)$ be a configuration of $\mathcal{G}_{L^{\prime}, r_{0}}$ with $\ell \in L_{\text {Min }}$. We construct $\sigma_{\mathrm{Min}}(\ell, \nu)$ as follows:

- if $\nu \geqslant r_{1}, \sigma_{\text {Min }}(\ell, \nu)=\sigma_{\text {Min }}^{0}(\ell, \nu)$;
- if $\nu<r_{1}, \ell \notin L^{\prime}$ and $\sigma_{\text {Min }}^{1}(\ell, \nu)=\left(t,\left(\ell, \ell^{f}\right)\right)$ for some delay $t$ (such that $\left.\nu+t \leqslant r_{1}\right)$, we let $\sigma_{\operatorname{Min}}(\ell, \nu)=\left(r_{1}-\nu+t^{\prime},\left(\ell, \ell^{\prime}\right)\right)$ where $\left(t^{\prime},\left(\ell, \ell^{\prime}\right)\right)=$ $\sigma_{\text {Min }}^{0}\left(\ell, r_{1}\right)$;
- otherwise $\sigma_{\text {Min }}(\ell, \nu)=\sigma_{\text {Min }}^{1}(\ell, \nu)$.

For all finite plays $\rho$ in $\mathcal{G}_{L^{\prime}, r_{0}}$ that conform to $\sigma_{\text {Min }}$, start in a configuration $(\ell, \nu)$ such that $\nu \in\left(r^{\prime \prime}, r_{0}\right]$ and $\ell \notin\left\{\ell^{\prime f} \mid \ell^{\prime} \in L\right\}$, and end in a final location, we show by induction that $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu)$. Note that $\rho$ either only contains valuations in $\left[r_{1}, r_{0}\right]$, or is of the form $(\ell, \nu) \xrightarrow{c}\left(\ell^{f}, \nu^{\prime}\right)$, or is of the form $(\ell, \nu) \xrightarrow{c} \rho^{\prime}$ with $\rho^{\prime}$ a run that satisfies the above restriction.

- If $\nu \in\left[r_{1}, r_{0}\right]$, then $\rho$ conforms with $\sigma_{\text {Min }}^{0}$, thus, as $\sigma_{\text {Min }}^{0}$ is fake-optimal, $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) \leqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)$ (the last inequality comes from the definition of $\left.r_{1}=\operatorname{left}_{L^{\prime}}\left(r_{0}\right)\right)$. Therefore, in the following cases, we assume that $\nu \in\left(r^{\prime \prime}, r_{1}\right)$.
- Consider then the case where $\rho$ is of the form $(\ell, \nu) \xrightarrow{c}\left(\ell^{f}, \nu^{\prime}\right)$.
- if $\ell \in L^{\prime} \cap L_{\mathrm{Min}}, \ell$ is urgent in $\mathcal{G}_{L^{\prime}, r_{0}}$, thus $\nu^{\prime}=\nu$. Furthermore, since $\rho$ conforms with $\sigma_{\text {Min }}$, by construction of $\sigma_{\text {Min }}$, the choice of $\sigma_{\text {Min }}^{1}$ on $\left(r^{\prime \prime}, r_{1}\right)$ consists in going to $\ell^{f}$, thus, as observed above, $(i) \wedge \neg(i i)$ holds for $\ell$. Therefore, $\operatorname{Val}_{\mathcal{G}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right)+\pi(\ell)\left(r_{0}-\nu\right)=$ $\varphi_{\ell_{f}}(\nu)=$ Price $_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho)$.
- If $\ell \in L_{\text {Min }} \backslash L^{\prime}$, by construction, it must be the case that $\sigma_{\text {Min }}(\ell, \nu)=$ $\left(r_{1}-\nu+t^{\prime},\left(\ell, \ell^{f}\right)\right)$ where $\left(t,\left(\ell, \ell^{f}\right)\right)=\sigma_{\text {Min }}^{1}(\ell, \nu)$ and $\left(t^{\prime},\left(\ell, \ell^{f}\right)\right)=$ $\sigma_{\text {Min }}^{0}\left(\ell, r_{1}\right)$. Thus, $\nu^{\prime}=r_{1}+t^{\prime}$. In particular, observe that $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho)=$ $\left(r_{1}-\nu\right) \pi(\ell)+\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{\prime}\right)$ where $\rho^{\prime}=\left(\ell, r_{1}\right) \xrightarrow{c^{\prime}}\left(\ell^{f}, \nu^{\prime}\right)$. As $\rho^{\prime}$ conforms with $\sigma_{\text {Min }}^{0}$ which is fake-optimal in $\mathcal{G}_{L^{\prime}, r_{0}}$, $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{\prime}\right) \leqslant$ $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\ell, r_{1}\right)=\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{1}\right)\left(\right.$ since $\left.r_{1}=\operatorname{left}\left(r_{0}\right)\right)$. Thus Price $\mathcal{G}_{\mathcal{L}^{\prime}, r_{0}}(\rho) \leqslant$ $\left(r_{1}-\nu\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{1}\right)=\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\rho^{\prime \prime}\right)$ where $\rho^{\prime \prime}=(\ell, \nu) \xrightarrow{c^{\prime \prime}}$ $\left(\ell^{f}, \nu+t\right)$ conforms with $\sigma_{\text {Min }}^{1}$ which is fake-optimal in $\mathcal{G}_{L^{\prime}, r_{1}}$. Therefore, $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) \leqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{1}}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)\left(\right.$ since $\left.r_{1}=\operatorname{left}\left(r_{0}\right)\right)$.
- If $\ell \in L_{\text {Max }}$ then $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho)=\left(\nu^{\prime}-\nu\right) \pi(\ell)+\varphi_{\ell_{f}}\left(\nu^{\prime}\right)=\left(\nu^{\prime}-\right.$ $\nu) \pi(\ell)+\left(r_{0}-\nu^{\prime}\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right)=\left(r_{0}-\nu\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right)$. By Lemma 14, since $\ell \in L_{\text {Max }} \backslash L_{u}$ ( $\ell$ is not urgent in $\mathcal{G}$ since $\ell^{f}$ exists), $\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{1}\right) \geqslant\left(r_{0}-r_{1}\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right)$. Furthermore, observe that if we define $\rho^{\prime}$ as the play $(\ell, \nu) \xrightarrow{c^{\prime}}\left(\ell^{f}, \nu\right)$ in $\mathcal{G}_{L^{\prime}, r_{1}}$, then $\rho^{\prime}$ conforms with $\sigma_{\text {Min }}^{1}$ and

$$
\begin{aligned}
\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\rho^{\prime}\right) & =\left(r_{1}-\nu\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{1}\right) \\
& \geqslant\left(r_{1}-\nu\right) \pi(\ell)+\left(r_{0}-r_{1}\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right) \\
& =\left(r_{0}-\nu\right) \pi(\ell)+\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{0}\right) \\
& =\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) .
\end{aligned}
$$

Thus, as $\sigma_{\text {Min }}^{1}$ is fake-optimal in $\mathcal{G}_{L^{\prime}, r_{1}}, \operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) \leqslant \operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\rho^{\prime}\right) \leqslant$ $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{1}}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)$.

- We finally consider the case where $\rho=(\ell, \nu) \xrightarrow{c} \rho^{\prime}$ with $\rho^{\prime}$ that starts in configuration $\left(\ell^{\prime}, \nu^{\prime}\right)$ such that $\ell^{\prime} \notin\left\{\ell^{\prime \prime f} \mid \ell^{\prime \prime} \in L\right\}$. By induction hypothesis $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{\prime}\right) \leqslant \operatorname{Val}_{\mathcal{G}}\left(\ell^{\prime}, \nu^{\prime}\right)$.
- If $\nu^{\prime} \leqslant r_{1}$, let $\rho^{\prime \prime}$ be the play of $\mathcal{G}_{L^{\prime}, r_{1}}$ starting in $\left(\ell^{\prime}, \nu^{\prime}\right)$ that conforms with $\sigma_{\text {Min }}^{1}$ and $\sigma_{\text {Max }}^{1}$. If $\rho^{\prime \prime}$ does not reach a final location, since $\sigma_{\text {Min }}^{1}$ is an NC-strategy, the prices of its prefixes tend to $-\infty$. By considering the strategy $\sigma_{\text {Min }}^{\prime}$ of Lemma 4, we would obtain a run conforming with $\sigma_{\text {Max }}^{1}$ of price smaller than $\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\ell^{\prime}, \nu^{\prime}\right)$ which would contradict the optimality of $\sigma_{\text {Max }}^{1}$. Hence, $\rho^{\prime \prime}$ reaches the target. Moreover, since $\sigma_{\text {Max }}^{1}$ is optimal and $\sigma_{\text {Min }}^{1}$ is fake-optimal, we finally know that $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\rho^{\prime \prime}\right)=\operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\ell^{\prime}, \nu^{\prime}\right)=\operatorname{Val}_{\mathcal{G}}\left(\ell^{\prime}, \nu^{\prime}\right)$ (since $\left.\nu^{\prime} \in\left[\operatorname{left}\left(r_{1}\right), r_{1}\right]\right)$. Therefore,

$$
\begin{aligned}
\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) & =\left(\nu^{\prime}-\nu\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime}\right)+\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{\prime}\right) \\
& \leqslant\left(\nu^{\prime}-\nu\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime}\right)+\operatorname{Val}_{\mathcal{G}}\left(\ell^{\prime}, \nu^{\prime}\right) \\
& =\left(\nu^{\prime}-\nu\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime}\right)+\operatorname{Price}\left(\rho^{\prime \prime}\right)=\operatorname{Price}\left((\ell, \nu) \xrightarrow{c^{\prime}} \rho^{\prime \prime}\right)
\end{aligned}
$$

Since the play $(\ell, \nu) \xrightarrow{c^{\prime}} \rho^{\prime \prime}$ conforms with $\sigma_{\text {Min }}^{1}$, we finally have $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) \leqslant \operatorname{Price}\left((\ell, \nu) \xrightarrow{c^{\prime}} \rho^{\prime \prime}\right) \leqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{1}}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}}(\ell, \nu)$.

- If $\nu^{\prime}>r_{1}$ and $\ell \in L_{\text {Max }}$, let $\rho^{1}$ be the play in $\mathcal{G}_{L^{\prime}, r_{1}}$ defined by $\rho^{1}=(\ell, \nu) \xrightarrow{c^{\prime}}\left(\ell^{f}, \nu\right)$ and $\rho^{0}$ the play in $\mathcal{G}_{L^{\prime}, r_{0}}$ defined by $\rho^{0}=$ $\left(\ell, r_{1}\right) \xrightarrow{c^{\prime \prime}} \rho^{\prime}$. We have

$$
\begin{aligned}
& \operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho)=\left(\nu^{\prime}-\nu\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime}\right)+\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{\prime}\right) \\
&=\underbrace{\varphi_{\ell_{f}}(\nu)}_{=\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{0}\right)}-\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{1}\right)+\underbrace{\left(\nu^{\prime}-r_{1}\right) \pi(\ell)+\pi\left(\ell, \ell^{\prime}\right)+\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{\prime}\right)} . \\
&=\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\rho^{1}\right)
\end{aligned}
$$

Since $\rho^{0}$ conforms with $\sigma_{\mathrm{Min}}^{0}$, fake-optimal, and reaches a final location, $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\rho^{0}\right) \leqslant \operatorname{Val}_{\mathcal{G}_{L^{\prime}, r_{0}}}\left(\ell, r_{1}\right)=\operatorname{Val}_{\mathcal{G}}\left(\ell, r_{1}\right)$ (since $r_{1}=$

```
Algorithm 2: solve(G)
    Input: SPTG \(\mathcal{G}=\left(L_{\text {Min }}, L_{\text {Max }}, L_{f}, L_{u}, \boldsymbol{\varphi}, \Delta, \pi\right)\)
    \(\boldsymbol{f}=\left(f_{\ell}\right)_{\ell \in L}:=\operatorname{solveInstant}(\mathcal{G}, 1) \quad / * f_{\ell}:\{1\} \rightarrow \overline{\mathbb{R}} * /\)
    \(r:=1\)
    while \(0<r\) do \(\quad / *\) Invariant: \(f_{\ell}:[r, 1] \rightarrow \overline{\mathbb{R}} * /\)
        \(\mathcal{G}^{\prime}:=\operatorname{wait}(\mathcal{G}, r, \boldsymbol{f}(r)) \quad / * r\)-SPTG
        \(\mathcal{G}^{\prime}=\left(L_{\mathrm{Min}}, L_{\mathrm{Max}}, L_{f}^{\prime}, L_{u}^{\prime}, \varphi^{\prime}, T^{\prime}, \pi^{\prime}\right) * /\)
        \(L_{u}^{\prime}:=L_{u}^{\prime} \cup L \quad / *\) every location is made urgent */
        \(b:=r\)
        repeat /* Invariant: \(f_{\ell}:[b, 1] \rightarrow \overline{\mathbb{R}} * /\)
            \(a:=\max \left(\operatorname{PossCP}_{\mathcal{G}^{\prime}} \cap[0, b)\right)\)
            \(\boldsymbol{x}=\left(x_{\ell}\right)_{\ell \in L}:=\operatorname{solveInstant}\left(\mathcal{G}^{\prime}, a\right) \quad / * x_{\ell}=\operatorname{Val}_{\mathcal{G}^{\prime}}(\ell, a) * /\)
            if \(\forall \ell \in L_{\text {Min }} \frac{f_{\ell}(b)-x_{\ell}}{b-a} \leqslant-\pi(\ell) \wedge \forall \ell \in L_{\text {Max }} \frac{f_{\ell}(b)-x_{\ell}}{b-a} \geqslant-\pi(\ell)\) then
                foreach \(\ell \in L\) do
                    \(f_{\ell}:=\left(\nu \in[a, b] \mapsto f_{\ell}(b)+(\nu-b) \frac{f_{\ell}(b)-x_{\ell}}{b-a}\right) \triangleright f_{\ell}\)
                    \(b:=a\); stop \(:=\) false
            else stop := true
        until \(b=0\) or stop
        \(r:=b\)
    return \(f\)
```

$\left.\operatorname{left}_{L^{\prime}}\left(r_{0}\right)\right)$. We also have that $\rho^{1}$ conforms with $\sigma_{\text {Min }}^{1}$, so the previous explanations already proved that $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{1}}}\left(\rho^{1}\right) \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu)$. As a consequence $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu)$.

- If $\nu^{\prime}>r_{1}$ and $\ell \in L_{\text {Min }}$, we know that $\ell$ is non-urgent, so that $\ell \notin L^{\prime}$. Therefore, by definition of $\sigma_{\operatorname{Min}}, \sigma_{\operatorname{Min}}(\ell, \nu)=\left(r_{1}-\nu+t^{\prime},\left(\ell, \ell^{\prime}\right)\right)$ where $\sigma_{\text {Min }}^{1}(\ell, \nu)=\left(t,\left(\ell, \ell^{f}\right)\right)$ for some delay $t$, and $\sigma_{\text {Min }}^{0}\left(\ell, r_{1}\right)=\left(t^{\prime},\left(\ell, \ell^{\prime}\right)\right)$. If we let $\rho^{1}$ be the play in $\mathcal{G}_{L^{\prime}, r_{1}}$ defined by $\rho^{1}=(\ell, \nu) \xrightarrow{c^{\prime}}\left(\ell^{f}, \nu\right)$ and $\rho^{0}$ the play in $\mathcal{G}_{L^{\prime}, r_{0}}$ defined by $\rho^{0}=\left(\ell, r_{1}\right) \xrightarrow{c^{\prime \prime}} \rho^{\prime}$, as in the previous case, we obtain that $\operatorname{Price}_{\mathcal{G}_{L^{\prime}, r_{0}}}(\rho) \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu)$.

As a consequence of this induction, we have shown that for all $\ell \in L$, and for all $\nu \in\left(r^{\prime \prime}, r_{1}\right)$, fake $\mathcal{G}_{\mathcal{G}_{L^{\prime}, r_{0}}}^{\sigma_{\text {Min }}}(\ell, \nu) \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu)$, which shows one inequality of (5), the other being obtained very similarly.

Algorithm 2 implements these ideas. Each iteration of the while loop computes a new game in the sequence $\mathcal{G}_{L \backslash L_{u}, 1}, \mathcal{G}_{L \backslash L_{u}, \text { left(1) }}, \ldots$ described above; solves it thanks to solveInstant; and thus computes a new portion of $\mathrm{Va} \mathrm{I}_{\mathcal{G}}$ on an interval on the left of the current point $r \in[0,1]$. More precisely, the vector $\left(\operatorname{Val}_{\mathcal{G}}(\ell, 1)\right)_{\ell \in L}$ is first computed in line 1. Then, the algorithm enters the while loop, and the game $\mathcal{G}^{\prime}$ obtained when reaching line 6 is $\mathcal{G}_{L \backslash L_{u}, 1}$. Then, the algorithm enters the repeat loop to analyse this game. Instead of building the whole value function of $\mathcal{G}^{\prime}$, Algorithm 2 builds only the parts of $\mathrm{Val}_{\mathcal{G}^{\prime}}$ that coincide with $\mathrm{Val}_{\mathcal{G}}$. It proceeds by enumerating the possible cutpoints $a$ of $\mathrm{Val}_{\mathcal{G}^{\prime}}$, starting in $r$, by decreasing valuations (line 8), and computes the value
of $\mathrm{Val}_{\mathcal{G}^{\prime}}$ in each cutpoint thanks to solveInstant (line 9), which yields a new piece of $\mathrm{Val}_{\mathcal{G}^{\prime}}$. Then, the if in line 10 checks whether this new piece coincides with $\mathrm{Val}_{\mathcal{G}}$, using the condition given by Proposition 13. If it is the case, the piece of $\mathrm{Val}_{\mathcal{G}^{\prime}}$ is added to $f_{\ell}$ (line 11); repeat is stopped otherwise. When exiting the repeat loop, variable $b$ has value left(1). Hence, at the next iteration of the while loop, $\mathcal{G}^{\prime}=\mathcal{G}_{L \backslash L_{u}, \text { left(1) }}$ when reaching line 6. By continuing this reasoning inductively, one concludes that the successive iterations of the while loop compute the sequence $\mathcal{G}_{L \backslash L_{u}, 1}, \mathcal{G}_{L \backslash L_{u}, \operatorname{left}(1)}, \ldots$ as announced, and rebuilds $\mathrm{Val}_{\mathcal{G}}$ from them. Termination in exponential time is ensured by Proposition 20 . each iteration of the while loop discovers at least one new cutpoint of $\mathrm{Val}_{\mathcal{G}}$, and there are at most exponentially many (note that a tighter bound on this number of cutpoints would entail a better complexity of our algorithm).

Example 1. Figure 11 shows the value functions of the SPTG of Figure 1. Here is how the algorithm Algorithm 2 obtains those functions. During the first iteration of the while loop, the algorithm computes the correct value functions until the cutpoint $\frac{3}{4}$ : in the repeat loop, at first $a=9 / 10$ but the slope in $\ell_{1}$ is smaller than the slope that would be granted by waiting, as depicted in Figure 1. Then, $a=3 / 4$ where the algorithm gives a slope of value -16 in $\ell_{2}$ while the price of this location of Max is -14 . During the first iteration of the while loop, the inner repeat loop thus ends with $r=3 / 4$. The next iterations of the while loop end with $r=\frac{1}{2}$ (because $\ell_{1}$ does not pass the test in line 10); $r=\frac{1}{4}$ (because of $\ell_{2}$ ) and finally with $r=0$, giving us the value functions on the entire interval $[0,1]$.

## 7. Towards more complex PTGs

In BLMR06, Rut11, HIJM13, general PTGs with non-negative prices are solved by reducing them to a finite sequence of SPTGs, by eliminating guards and resets. It is thus natural to try and adapt these techniques to our general case, in which case Algorithm 2 would allow us to solve general PTGs with arbitrary prices. Let us explain where are the difficulties of such a generalisation. The technique used to remove strict guards from the transitions of the PTGs, i.e. guards of the form $(a, b],[b, a)$ or $(a, b)$ with $a, b \in \mathbb{N}$, consists in enhancing the locations with regions while keeping an equivalent game. This technique can be adapted to arbitrary weights.

Formally, let $\mathcal{G}=\left(L_{\text {Min }}, L_{\text {Max }}, L_{f}, L_{u}, \boldsymbol{\varphi}, \Delta, \pi\right)$ be a PTG. We define the region-PTG of $\mathcal{G}$ as $\mathcal{G}^{\prime}=\left(L_{\text {Min }}^{\prime}, L_{\text {Max }}^{\prime}, L_{f}^{\prime}, L_{u}^{\prime}, \varphi^{\prime}, \Delta^{\prime}, \pi^{\prime}\right)$ where:

- $L_{\text {Min }}^{\prime}=\left\{(\ell, I) \mid \ell \in L_{\text {Min }}, I \in \operatorname{Reg}_{\mathcal{G}}\right\} ;$
- $L_{\text {Max }}^{\prime}=\left\{(\ell, I) \mid \ell \in L_{\text {Max }}, I \in \operatorname{Reg}_{\mathcal{G}}\right\} ;$
- $L_{f}=\left\{(\ell, I) \mid \ell \in L_{f}, I \in \operatorname{Reg}_{\mathcal{G}}\right\} ;$
- $L_{u}=\left\{(\ell, I) \mid \ell \in L_{u}, I \in \operatorname{Reg}_{\mathcal{G}}\right\} ;$
- $\forall(\ell, I) \in L_{f}^{\prime}, \varphi_{\ell, I}^{\prime}=\varphi_{\ell}$;


Figure 11: Value functions of the SPTG of Figure 1

$$
\begin{aligned}
\Delta^{\prime}= & \left\{\left((\ell, I), \overline{I_{g} \cap I}, R,\left(\ell^{\prime}, I^{\prime}\right)\right) \mid\left(\ell, I_{g}, R, \ell^{\prime}\right) \in \Delta, I^{\prime}=\left\{\begin{array}{ll}
I & \text { if } R=\perp \\
\{0\} & \text { otherwise }
\end{array}\right\}\right. \\
& \cup\left\{\left(\left(\ell,\left(M_{k}, M_{k+1}\right)\right),\left\{M_{k+1}\right\}, \perp,\left(\ell,\left\{M_{k+1}\right\}\right)\right) \mid \ell \in L,\left(M_{k}, M_{k+1}\right) \in \operatorname{Reg}_{\mathcal{G}}\right\} \\
& \cup\left\{\left(\left(\ell,\left\{M_{k}\right\}\right),\left\{M_{k}\right\}, \perp,\left(\ell,\left(M_{k}, M_{k+1}\right)\right)\right) \mid \ell \in L,\left(M_{k}, M_{k+1}\right) \in \operatorname{Reg}_{\mathcal{G}}\right\} ;
\end{aligned}
$$

Transitions belonging to the last two sets are called waiting transition denoted by WaitTr.

- $\forall(\ell, I) \in L^{\prime}, \pi^{\prime}(\ell, I)=\pi(\ell)$; and $\forall\left((\ell, I), I_{g}, R,\left(\ell^{\prime}, I^{\prime}\right)\right) \in \Delta^{\prime}$, if $\left(\ell, I_{g}, R, \ell^{\prime}\right) \in$ $\Delta$, then $\pi\left((\ell, I), I_{g}, R,\left(\ell^{\prime}, I^{\prime}\right)\right)=\pi\left(\ell, I_{g}, R, \ell\right)$, else $\pi\left((\ell, I), I_{g}, R,\left(\ell^{\prime}, I^{\prime}\right)\right)=$ 0.

It is easy to verify that, in all configurations $\left(\left(\ell,\left\{M_{k}\right\}\right), \nu\right)$ reachable from the null valuation, the valuation $\nu$ is $M_{k}$. More interestingly, in all configurations $\left(\left(\ell,\left(M_{k}, M_{k+1}\right)\right), \nu\right)$ reachable from the null valuation, the valuation $\nu$ is in
21. Let $\mathcal{G}$ be a $P T G$, and $\mathcal{G}^{\prime}$ be its region- $P G$ defined as before. For $(\ell, I) \in L \times \operatorname{Reg}_{\mathcal{G}}$ and $\nu \in I, \operatorname{Val}_{\mathcal{G}}(\ell, \nu)=\operatorname{Val}_{\mathcal{G}^{\prime}}((\ell, I), \nu)$. Moreover, we can transform an $\varepsilon$-optimal strategy of $\mathcal{G}^{\prime}$ into an $\varepsilon^{\prime}$-optimal strategy of $\mathcal{G}$ with $\varepsilon^{\prime}<2 \varepsilon$ and vice-versa.

Proof. The proof consists in replacing strategies of $\mathcal{G}^{\prime}$ where players can play the plays of $\mathcal{G} . g$ is defined by induction on the length $n$ of the plays so that for a play $\rho$ of length $n$ we have (1) $|\operatorname{Price}(\rho)-\operatorname{Price}(g(\rho))| \leqslant 2 \Pi^{\mathrm{loc}}\left(1-\frac{1}{2^{n}}\right) \varepsilon$ and (2) there exists $\ell \in L$ and $I \in \operatorname{Reg}_{\mathcal{G}}$ such that $g(\rho)$ and $\rho$ ends in the states $\ell$ and $(\ell, I)$ and their valuations are both in $I$ and differ of at most $\frac{1}{2^{n+1}} \varepsilon$.

If $n=0$, let $\rho=((\ell, I), \nu)$ be a play of $\mathcal{G}^{\prime}$ of length 0 , then $g(\rho)=\left(\ell, \nu^{\prime}\right)$, where $\nu^{\prime}=\nu \pm \frac{\varepsilon}{2}$ if $I$ is not an interval and $\nu$ is an endpoint of $I$, and $\nu^{\prime}=\nu$ otherwise (so that $\nu^{\prime} \in I$ in every case).

For $n>0$, we suppose $g$ defined on every play of length at most $n$ which does not end with a waiting transition. Let $\rho=\left(\left(q_{1}, I_{1}\right), \nu_{1}\right) \xrightarrow{t_{1}, \delta_{1}, c_{1}} \ldots \xrightarrow{t_{n}, \delta_{n}, c_{n}}$ $\left(\left(q_{n}, I_{n}\right), \nu_{n}\right) \xrightarrow{t_{n+1}, \delta_{n+1}, c_{n+1}}\left(\left(q_{n+1}, I_{n+1}\right), \nu_{n+1}\right)$ with $\delta_{n+1} \notin$ WaitTr. Let last $=\max \left(\left\{k \leqslant n \mid t r_{k} \notin\right.\right.$ WaitTr $\left.\}\right)($ with $\max \emptyset=0)$. Then, by induction, there exists $\rho^{\prime}=\left(q_{1}, \nu_{1}\right) \rightarrow \cdots \rightarrow\left(q_{\text {last }+1}, v_{\text {last }+1}^{\prime}\right)$ such that $g\left(\rho_{\text {last }}\right)=\rho^{\prime}$ (where $\rho_{\mid \text {last }}$ is the prefix of length last of $\rho$ ), $\mid \operatorname{Price}\left(\rho_{\mid \text {last }}\right)$ - Price $\left(g\left(\rho_{\mid \text {last }}\right)\right) \mid \leqslant$ $2 \Pi^{\text {loc }}\left(1-\frac{1}{2^{\text {last }}}\right) \varepsilon$ and $\left|\nu_{\text {last }+1}^{\prime}-\nu_{\text {last }+1}\right| \leqslant \frac{1}{2^{\text {last }+1}} \varepsilon$. Then we choose $g(\rho)=$ $\rho^{\prime} \xrightarrow{t, \delta_{n+1}, c}\left(q_{n+1}, \nu_{n+1}^{\prime}\right)$ where

- if $\delta_{n+1}$ is enabled in $\mathcal{G}$ in $\left(q_{\text {last }+1}=q_{n}, \nu_{n}+t_{n+1}\right), t=\nu_{n}+t_{n+1}-\nu_{\text {last }}^{\prime}$;
- otherwise, as the permissive interval of $\mathcal{G}^{\prime}$ are the closure of the permissive interval of $\mathcal{G}$, then there exists $z \in\{+1 ;-1\}$ such that for $t=\nu_{n}+t_{n+1}-$ $\nu_{\text {last }}^{\prime}+\frac{z \varepsilon}{2^{n+2}}, \delta_{n+1}$ is enabled in $\mathcal{G}$ and $\nu_{\text {last }}^{\prime}+t$ and $\nu_{n}+t_{n+1}$ belong to the same region.

Thus, in both cases, $\left|\nu_{n+1}-\nu_{n+1}^{\prime}\right| \leqslant \frac{\varepsilon}{2^{n+2}}$ and $\nu_{n+1} \neq \nu_{n+1}^{\prime}$ iff $I$ is not a singleton, $\nu_{n+1}$ is on a border, $\nu_{n+1}^{\prime}$ is close to this border and $\delta_{n+1}$ does not contain a reset. Moreover,

$$
\begin{aligned}
|\operatorname{Price}(\rho)-\operatorname{Price}(g(\rho))|= & \left|\operatorname{Price}\left(\rho_{\mid \text {last }}\right)+\left(\nu_{n+1}-\nu_{\text {last }}\right) \pi\left(q_{\text {last }}\right)+\pi\left(\delta_{n+1}\right)-\operatorname{Price}(g(\rho))\right| \\
\leqslant & \left|\operatorname{Price}\left(\rho_{\mid \text {last }}\right)-\operatorname{Price}\left(g\left(\rho_{\mid \text {last }}\right)\right)\right| \\
& +\left|\left(\nu_{n+1}-\nu_{\text {last }}\right) \pi\left(q_{\text {last }}\right)+\pi\left(\delta_{n+1}\right)+\operatorname{Price}\left(g\left(\rho_{\mid \text {last }}\right)\right)-\operatorname{Price}(g(\rho))\right| \\
\leqslant & 2 \Pi^{\text {loc }}\left(1-\frac{1}{2^{\text {last }}}\right) \varepsilon+\left|\left(\nu_{\text {last }}^{\prime}-\nu_{\text {last }}\right) \pi\left(q_{\text {last }}\right)+\left(\nu_{n+1}-\nu_{n+1}^{\prime}\right) \pi\left(q_{\text {last }}\right)\right| \\
\leqslant & 2 \Pi^{\mathrm{loc}}\left(1-\frac{1}{2^{\text {last }}}\right) \varepsilon+\left|\frac{\varepsilon}{2^{\text {last }+1}} \pi\left(q_{\text {last }}\right)\right|+\left|\frac{\varepsilon}{2^{n+2}} \pi\left(q_{\text {last }}\right)\right| \\
\leqslant & 2 \Pi^{\text {loc }}\left(1-\frac{1}{2^{\text {last }}}\right) \varepsilon+\frac{\Pi^{\text {loc }} \varepsilon}{2^{\text {last }+1}}+\frac{\Pi^{\text {loc }} \varepsilon}{2^{n+2}} \\
\leqslant & 2 \Pi^{\text {loc }}\left(1-\frac{1}{2^{\text {last }+1}}\right) \varepsilon \\
\leqslant & 2 \Pi^{\text {loc }}\left(1-\frac{1}{2^{\text {n+1 }}}\right) \varepsilon .
\end{aligned}
$$

Let $\sigma_{\text {Min }}$ be a strategy of $\operatorname{Min}$ in $\mathcal{G}$. Using the transformation $g$, we will build by induction a strategy $\sigma_{\text {Min }}^{\prime}$ in $\mathcal{G}^{\prime}$ such that for all plays $\rho$ whose last transition does not belong to WaitTr and conforming with $\sigma_{\text {Min }}^{\prime}, g(\rho)$ conforms with $\sigma_{\text {Min }}$.

Let $\rho$ be a play of $\mathcal{G}^{\prime}$ whose last transition does not belong to WaitTr such that $g(\rho)$ conforms with $\sigma_{\text {Min }}$ (which is the case of all plays of length 0 ). $\rho$ and $g(\rho)$ ends in the locations $(q, I)$ and $q$ respectively.

- If $\rho$ ends in a configuration of Max, then the choice does not depend on $\sigma_{\text {Min }}$ or $\sigma_{\text {Min }}^{\prime}$. Let $(t, \delta)$ be a choice of Max in $\mathcal{G}^{\prime}$ with price $c$. If $\delta$ belongs to WaitTr, then the new configuration also belongs to Max where he will make another choice. Let $\rho^{\prime}$ be the extension of $\rho$ until the first transition $\delta^{\prime}$ such that $\delta^{\prime} \notin$ WaitTr. $g\left(\rho^{\prime}\right)$ conforms with $\sigma_{\text {Min }}$ as the configuration where $g(\rho)$ ends is controlled by Max and $g\left(\rho^{\prime}\right)$ only has one more transition than $g(\rho)$.
- If $\rho$ ends in configuration of Min, then there exists $t, \delta, c, q^{\prime}, \nu^{\prime}$ such that $g(\rho) \xrightarrow{t, \delta, c}\left(q^{\prime}, \nu^{\prime}\right)$ conforms with $\sigma_{\text {Min }}$. As taking a waiting transition does not change the ownership of the configuration, we consider here multiple successive choices of Min as one choice: $\sigma_{\text {Min }}^{\prime}(\rho)$ is such that $\rho^{\prime}=\rho \xrightarrow{t_{1}, \delta_{1}, c_{1}}$ $\ldots \xrightarrow{t_{k}, \delta_{k}, c_{k}}\left(\left(q, I^{\prime \prime}\right), \nu\right) \xrightarrow{t_{k+1}, \delta, c_{k+1}}\left(\left(q^{\prime}, I^{\prime}\right), \nu^{\prime}\right)$ where $\forall i \leqslant k, \delta_{i} \in$ WaitTr conforms with $\sigma_{\text {Min }}^{\prime}$. This is possible as if $\delta$ is allowed in a configuration $(q, \nu)$ in $\mathcal{G}$ then it is allowed too in a configuration $((q, I), \nu)$ with the appropriate $I$. Then $g\left(\rho^{\prime}\right)=g(\rho) \xrightarrow{\delta, t r, c}\left(q^{\prime}, \nu^{\prime}\right)$, thus $g\left(\rho^{\prime}\right)$ conforms with $\sigma_{\text {Min }}$.

As no accepting plays of $\mathcal{G}^{\prime}$ end with a transition of WaitTr, every accepting play $\rho$ conforming with $\sigma_{\text {Min }}^{\prime}$ verifies that $g(\rho)$ conforms with $\sigma_{\text {Min }}$. Thus for every configuration $s$, $\operatorname{Price}_{\mathcal{G}^{\prime}}\left(s, \sigma_{\text {Min }}^{\prime}\right) \leqslant \operatorname{Price}_{\mathcal{G}}\left(s, \sigma_{\text {Min }}\right)+2 \Pi^{\text {loc }} \varepsilon$. Therefore $\operatorname{Val}_{\mathcal{G}^{\prime}}(s) \leqslant \operatorname{Val}_{\mathcal{G}}(s)$.

Reciprocally, let $\sigma_{\text {Min }}^{\prime}$ be a strategy of $\operatorname{Min}$ in $\mathcal{G}^{\prime}$. We will now build by induction a strategy $\sigma_{\operatorname{Min}}$ in $\mathcal{G}$ such that for all plays $\rho$ conforming with $\sigma_{\text {Min }}$, there exists a play in $g^{-1}(\rho)$ that conforms with $\sigma_{\text {Min }}^{\prime}$.

Let $\rho$ be a play of $\mathcal{G}$ conforming with $\sigma_{\text {Min }}$ such that there exists $\rho^{\prime} \in g^{-1}(\rho)$ conforming with $\sigma_{\text {Min }}^{\prime}$ (which is the case of all plays of length 0 ). $\rho^{\prime}$ and $\rho$ ends in the configuration $\left((q, I), \nu^{\prime}\right)$ and $(q, \nu)$.

- If $\rho$ ends in configuration of Max, then the choice does not depend on $\sigma_{\text {Min }}$ or $\sigma_{\text {Min }}^{\prime}$. Let $(t, \delta)$ be a choice of Max in $\mathcal{G}$ with price $c$ and let $\tilde{\rho}$ be the extension of $\rho$ by this choice. There exists $\left(t_{1}, \delta_{1}, c_{1}\right), \ldots,\left(t_{k+1}, \delta_{k+1}, c_{k+1}\right)$ such that $\forall i \leqslant k, \delta_{i} \in$ WaitTr, $\delta_{k+1}=\delta$ and $\sum_{i=1}^{k+1} t_{i}=\nu+t-\nu^{\prime}$. Let $\rho_{c}=\rho^{\prime} \xrightarrow{t_{1}, \delta_{1}, c_{1}} \ldots \xrightarrow{t_{k}, \delta_{k}, c_{k}}\left(\left(q, I^{\prime \prime}\right), \nu_{k}\right) \xrightarrow{t_{k+1}, \delta, c_{k+1}}\left(\left(q^{\prime}, I^{\prime}\right), \nu_{k+1}\right)$, then $\rho_{c}$ conforms with $\sigma_{\text {Min }}^{\prime}$ (as Min did not take a single decision) and $g\left(\rho_{c}\right)=\tilde{\rho}$.
- If $\rho$ ends in a configuration of Min, then there exists a play $\rho_{c}=\rho \xrightarrow{t_{1}, \delta_{1}, c_{1}}$ $\ldots \xrightarrow{t_{k}, \delta_{k}, c_{k}}\left(\left(q, I^{\prime \prime}\right), \nu_{k}\right) \xrightarrow{t_{k+1}, \delta, c_{k+1}}\left(\left(q^{\prime}, I^{\prime}\right), \nu_{k+1}\right)$ such that $\rho_{c}$ conforms with $\sigma_{\text {Min }}^{\prime}$. We choose $\sigma_{\text {Min }}(\rho)=(t, \delta)$ such that for the adequate price $c$, $g\left(\rho_{c}\right)=\rho \xrightarrow{t, \delta, c}\left(q^{\prime}, v^{\prime \prime}\right)$. This is possible as $t+\nu^{\prime} \in I^{\prime \prime}$.

Every accepting play $\rho$ conforming with $\sigma_{\text {Min }}$ verifies $\exists \rho^{\prime} \in g^{-1}(\rho)$ conforming with $\sigma_{\text {Min }}$. Thus for every configuration $s, \operatorname{Price}_{\mathcal{G}}\left(s, \sigma_{\text {Min }}\right) \leqslant \operatorname{Price}_{\mathcal{G}^{\prime}}\left(s, \sigma_{\text {Min }}\right)+$ $2 \Pi^{\mathrm{loc}} \varepsilon$. Therefore $\operatorname{Val}_{\mathcal{G}^{\prime}}(s) \geqslant \operatorname{Val}_{\mathcal{G}}(s)$. Hence $\operatorname{Val}_{\mathcal{G}^{\prime}}(s)=\operatorname{Val}_{\mathcal{G}}(s)$.

The technique to handle resets, however, consists in bounding the number of clock resets that can occur in each play following an optimal strategy of Min or Max. Then, the PTG can be unfolded into a reset-acyclic PTG with the same value. By reset-acyclic, we mean that no cycle in the configuration graph visits a transition with a reset. This reset-acyclic PTG can be decomposed into a finite number of components that contain no reset and are linked by transitions with resets. These components can be solved iteratively, from the bottom to the top, turning them into SPTGs. Thus, if we assume that the PTGs we are given as input are reset-acyclic, we can solve them in exponential time, and show that their value functions are cost functions with at most exponentially many cutpoints, using our techniques. In BLMR06 the authors showed that with one-clock PTG and positive prices only we could bound the number of reset by $n$, the number of states, without changing the value functions. Unfortunately, the arguments to bound the number of resets do not hold for arbitrary prices, as shown by the PTG in Figure 12 . We claim that $\operatorname{Val}\left(\ell_{0}\right)=0$; that Min has no optimal strategy, but a family of $\varepsilon$-optimal strategies $\sigma_{\text {Min }}^{\varepsilon}$ each with value $\varepsilon$; and that each $\sigma_{\text {Min }}^{\varepsilon}$ requires memory whose size depends on $\varepsilon$ and might yield a play visiting at least $1 / \varepsilon$ times the reset between $\ell_{0}$ and $\ell_{1}$ (hence the number of resets cannot be bounded). For all $\varepsilon>0, \sigma_{\text {Min }}^{\varepsilon}$ consists in: waiting $1-\varepsilon$ time units in $\ell_{0}$, then going to $\ell_{1}$ during the $\lceil 1 / \varepsilon\rceil$ first visits to $\ell_{0}$; and to go directly to $\ell_{f}$ afterwards. Against $\sigma_{\text {Min }}^{\varepsilon}$, Max has two possible choices:
(i) either wait 0 time unit in $\ell_{1}$, wait $\varepsilon$ time units in $\ell_{2}$, then reach $\ell_{f}$; or


Figure 12: A PTG where the number of resets in optimal plays cannot be bounded a priori.
(ii) wait $\varepsilon$ time unit in $\ell_{1}$ then force the cycle by going back to $\ell_{0}$ and wait for Min's next move.

Thus, all plays according to $\sigma_{\text {Min }}^{\varepsilon}$ will visit a sequence of locations which is either of the form $\ell_{0}\left(\ell_{1} \ell_{0}\right)^{k} \ell_{1} \ell_{2} \ell_{f}$, with $0 \leqslant k<\lceil 1 / \varepsilon\rceil$; or of the form $\ell_{0}\left(\ell_{1} \ell_{0}\right)^{\left\lceil\frac{1}{\varepsilon}\right.} \ell_{f}$. In the former case, the price of the play will be $-k \varepsilon+0+\varepsilon=-(k-1) \varepsilon \leqslant \varepsilon$; in the latter, $-\varepsilon(\lceil 1 / \varepsilon\rceil)+1 \leqslant 0$. This shows that $\operatorname{Val}\left(\ell_{0}\right)=0$, but there is no optimal strategy as none of these strategies allow one to guarantee a price of 0 (neither does the strategy that waits 1 time unit in $\ell_{0}$ ).

If bounding the number of resets is not possible in the general case, it could be done if one adds constraints on the cycles of the game. This kind of restriction was used in BCR14 where they introduce the notion of robust games. Such games requires among other things that there exists $\kappa>0$ such that every play starting and ending in the same pair location and time region has either a positive price or a price smaller than $-\kappa$. Here we require a less powerful assumption as we put this restriction only on cycles containing a reset.

Definition 5. Given $\kappa>0$, a $\kappa$-negative-reset-acyclic $\operatorname{PTG}$ (NRAPTG) is a $P T G$ where for every state $\ell \in L$ and every cyclic finite play $\rho$ starting and ending in $(\ell, 0)$, either $\operatorname{Price}(\rho) \geqslant 0$ or $\operatorname{Price}(\rho)<-\kappa$.

The PTG of Figure 12 is not a $\kappa$-NRAPTG for any $\kappa>0$ as the play $\left(\ell_{0}, 0\right) \xrightarrow{0}\left(\ell_{1}, 1-\kappa / 2\right) \xrightarrow{-\kappa / 2}\left(\ell_{0}, 0\right)$ is a cycle containing a reset and with a negative price strictly greater than $-\kappa$. On the contrary, in Figure 13 we show a -1 -NRAPTG and its region PTG. Here, every cycle containing a reset is between $\ell_{0}$ and $\ell_{1}$ and such cycles have at most price -1 . The value of this PTG is 0 but no strategies for Max can achieve it because of the guard $x>0$. As this guard is not strict anymore in the region PTG, both player have an optimal strategy in this game (this is not always the case).

In order to bound the number of resets of a $\kappa$-NRAPTG we first prove a bound on the value of such games, that will be useful in the following. We let $k=\left|\operatorname{Reg}_{\mathcal{G}}\right|$ be the number of regions.

Lemma 22. For all $\kappa$-NRAPTGs $\mathcal{G}$, for all $(\ell, \nu) \in \operatorname{Conf}_{\mathcal{G}}$ : either $\operatorname{Val}_{\mathcal{G}}(\ell, \nu) \in$ $\{-\infty,+\infty\}$, or

$$
-|L| M \Pi^{\mathrm{loc}}-|L|^{2}(|L|+2) \Pi^{\mathrm{tr}} \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu) \leqslant|L| M \Pi^{\mathrm{loc}}+|L| k \Pi^{\mathrm{tr}} .
$$

Proof. Consider the case where $\operatorname{Val}_{\mathcal{G}}(\ell, \nu) \notin\{-\infty,+\infty\}$. Let $\kappa>2 \varepsilon>0$. Then, there exist $\sigma_{\text {Min }}$ and $\sigma_{\text {Max }} \varepsilon$-optimal strategies for Min and Max, respectively.

Let $\sigma_{\text {Min }}^{c}$ be any memoryless strategy of Min in the reachability timed game induced by $\mathcal{G}$ such that no play consistent with $\sigma_{\operatorname{Min}}^{\mathcal{c}}$ goes twice in the same couple (location, region). If such a strategy does not exist, as the clock constraints are


Figure 13: A -1-NRAPTG and its region PTG (some guards were removed for a better readability)
the same during the first and second occurrences of this couple, Max can enforce the cycle infinitely often, thus the reachability game is winning for him and the value of $\mathcal{G}$ is $+\infty$. Let us note $\rho=\operatorname{Play}\left((\ell, \nu), \sigma_{\operatorname{Min}}^{\neg}, \sigma_{\mathrm{Max}}\right)$. By $\varepsilon$-optimality of $\sigma_{\mathrm{Max}}$, $\operatorname{Price}(\rho) \geqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu)-\varepsilon$. Let $\operatorname{Price}{ }^{\operatorname{tr}}(\rho)$ be the price of $\rho$ due to the prices of the transitions, and Price ${ }^{\text {loc }}(\rho)$ be the price due to the time elapsed in the locations of the game: $\operatorname{Price}(\rho)=\operatorname{Price}^{\operatorname{tr}}(\rho)+\operatorname{Price}^{\text {loc }}(\rho)$. As there are no cycles in the game according to couples (location, region), there are at most $|L| k$ transitions, thus Price ${ }^{\operatorname{tr}}(\rho) \leqslant|L| k \Pi^{\mathrm{tr}}$. Moreover, the absence of cycles also implies that we do not take two transitions with a reset ending in the same location or one transition with a reset ending in the initial location, thus we take at most $|L|-1$ such transitions. Therefore at most $|L| M$ units of time elapsed and $\operatorname{Price}^{\text {loc }}(\rho) \leqslant|L| M \Pi^{\text {loc }}$. This implies that

$$
\operatorname{Val}_{\mathcal{G}}(\ell, \nu)-\varepsilon \leqslant \operatorname{Price}(\rho) \leqslant|L| M \Pi^{\mathrm{loc}}+|L| k \Pi^{\operatorname{tr}}
$$

By taking the limit of $\varepsilon$ towards 0 , we obtain the announced upper bound.
We now prove the lower bound on the value. To that extent, consider now
We want to lower bound the prince of $\rho$, therefore non-negative cycles can be safely ignored. Let us show that there are no negative cycles around a transition with a reset. If it was the case, since the game is a $\kappa$-NRAPTG, this cycle has weight at most $-\kappa$. Since the strategy $\sigma_{\text {Max }}$ is $\varepsilon$-optimal, and $\kappa>\varepsilon$, it is not the capability to enforce this cycle, and to exit it (otherwise, Max would keep him inside to get value $+\infty$ ): but then, Min could decide to cycle as long as he wants, then guaranteeing a value as low as possible, which contradicts the fact that $\operatorname{Val}(\ell, \nu) \notin\{-\infty,+\infty\}$. Therefore, the only cycles in $\rho$ around transitions with resets, are non-negative cycles. This implies that its price is bounded below by the price of a subplay obtained by removing the cycles in $\rho$.

We now consider a play where each reset transition is taken at most once in $\rho$, and lower-bound its price.

If $\rho$ contains a cycle around a location $\ell^{\prime} \in L_{\text {Max }}$ without reset transitions, this cycle has the form $\left(\ell^{\prime}, \nu^{\prime}\right) \xrightarrow{c^{\prime}}\left(\ell^{\prime \prime}, \nu^{\prime}+t\right) \cdots \xrightarrow{c^{\prime \prime}}\left(\ell^{\prime}, \nu^{\prime \prime}\right)$ with $\nu^{\prime \prime} \geqslant \nu^{\prime}$, followed in $\rho$ by a transition towards configuration $\left(\ell^{\prime \prime \prime}, \nu^{\prime \prime}+t^{\prime}\right)$. Thus, another strategy for Max could have consisted in skipping the cycle by choosing as delay in the first location $\ell^{\prime}, \nu^{\prime \prime}-\nu^{\prime}+t^{\prime}$ instead of $t$. This would get a new strategy that cannot make the price increase above $\mathrm{Val}_{\mathcal{G}}(\ell, \nu)+\varepsilon$, since it is still playing against an $\varepsilon$-optimal strategy of Min. Therefore, we can consider the subplay $\rho_{f}$ of $\rho$ where all such cycles are removes: we still have $\operatorname{Price}\left(\rho_{f}\right) \leqslant \operatorname{Val}_{\mathcal{G}}(\ell, \nu)+\varepsilon$.

Suppose now that $\rho_{f}$ contains a cycle around a location $\ell^{\prime} \in L_{\text {Min }}$ without reset transitions, of the form $\left(\ell^{\prime}, \nu^{\prime}\right) \xrightarrow{c^{\prime}}\left(\ell^{\prime \prime}, \nu^{\prime}+t\right) \cdots \xrightarrow{c^{\prime \prime}}\left(\ell^{\prime}, \nu^{\prime \prime}\right)$ with $\nu$ and $\nu^{\prime}$ in the same region, composed of Min's locations only, and followed in $\rho$ by a transition towards configuration ( $\ell^{\prime \prime \prime}, \nu^{\prime \prime}+t^{\prime}$ ). Then, the transition price of this cycle is non-negative, otherwise Min could enforce this cycle he entirely controls, while letting only a bounded time pass (smaller and smaller as the number of cycles grow). This is not possible.

Therefore, we have that two occurrences of a same Max's location in $\rho_{f}$ are separated by a reset transition and two occurrences of a same Min's couple (location,region) are either separated by a reset or by a Max's location. As there is at most $|L|-1$ resets, $|L|$ locations of Max and $|L| k$ couples (location,region) for Min, $\rho_{t}$ contains at most $|L|^{2}$ states of Max and $|L| k\left(|L|^{2}+|L|-1+1\right)$ locations of Min, which makes for at most $|L|^{2}(|L| k+k+1)$ locations. Thus Price ${ }^{\text {loc }}\left(\rho_{t}\right) \geqslant-|L|^{2}(|L| k+k+1) \Pi^{\text {loc. }}$. Moreover, as at most $|L|-1$ resets are taken in $\rho_{f}$ and that the game is bounded by $M$, Price ${ }^{\text {loc }}\left(\rho_{f}\right) \geqslant-|L| M \Pi^{\text {loc }}$. This implies that
$\operatorname{Val}_{\mathcal{G}}(\ell, \nu)+\varepsilon \geqslant \operatorname{Price}^{\mathrm{loc}}\left(\rho_{f}\right)+\operatorname{Price}^{\mathrm{tr}}\left(\rho_{t}\right) \geqslant-|L| M \Pi^{\mathrm{loc}}-|L|^{2}(|L| k+k+1) \Pi^{\mathrm{tr}}$.
Taking the limit when $\varepsilon$ tends to 0 , we obtain the desired lower bound.
Using this bound on the value of a $\kappa$-NRAPTG one can give a bound on the number of cycles needed to be allowed. The idea is that if a reset is taken twice then if the generated cycle has positive price, either Min can modify its strategy so that it does not take this cycle or the value of the game is $+\infty$ as Max can stop Min to reach an accepting state. On the contrary if the cycle has negative price, then by definition of a $\kappa$-NRAPTG, this price is less than $-\kappa$. Thus by allowing enough such cycles, as we have bounds on the values of the game, we know when we will have enough cycles to get under the lower bound of the value of the game. By solving the copies of the game, if we reach a value that is smaller than the lower bound of the value, then it means that the value is $-\infty$.

Lemma 23. For all $\kappa>0$, the value of a $\kappa$-NRAPTG can be computed by solving $k=\left\lceil\kappa \times 2 n\left(\nu^{\text {sup }}-\nu_{\text {inf }}\right)\right\rceil$ PTGs without resets and using the same set of guards where $\nu^{\text {sup }}$ and $\nu_{\text {inf }}$ are the upper and lower bound of the value of the game given by Lemma 22. Moreover, from $\varepsilon$-optimal strategies on those $k$ games, we can build $k \varepsilon$-optimal strategies in the original game.

With this, we can conclude:
Theorem 24. Let $\kappa>0$ and $\mathcal{G}$ be a $\kappa$-NRAPTG. Then for every location $q \in Q$, the function $v \mapsto \operatorname{Val}_{\mathcal{G}}(q, \nu)$ is computable in EXPTIME and are piecewiseaffine functions with at most an exponential number of cutpoints. Moreover, for
every $\varepsilon>0$, there exists (and we can effectively compute) $\varepsilon$-optimal strategies for both players.

The robust games defined in BCR14 restricted to one-clock are a subset of the NRAPTG, therefore their value is computable with the same complexity. While we cannot extend the computation of the value to all (one-clock) PTGs, we can still obtain information on the nature of the value function:

Theorem 25. The value functions of all one-clock PTGs are cost functions with at most exponentially many cutpoints.

Proof. Let $\mathcal{G}$ be a one-clock PTG. Let us replace all transitions ( $\ell, g, \top, \ell^{\prime}$ ) resetting the clock by $\left(\ell, g, \perp, \ell^{\prime \prime}\right)$, where $\ell^{\prime \prime}$ is a new final location with $\varphi_{\ell^{\prime \prime}}=$ $\operatorname{Val}_{\mathcal{G}}(\ell, 0)$-observe that $\mathrm{Va}_{\mathcal{G}}(\ell, 0)$ exists even if we cannot compute it, so this transformation is well-defined. This yields a reset-acyclic PTG $\mathcal{G}^{\prime}$ such that $\mathrm{Val}_{\mathcal{G}^{\prime}}=\mathrm{Val}_{\mathcal{G}}$.

## 8. Conclusion

In this work, we study, for the first time, priced timed games with arbitrary weights and one clock, showing how to compute optimal values and strategies in exponential time for the special case of simple games. This complexity result is comparable with previously obtained results in the case of non-negative weights only [HIJM13, Rut11, even though we follow different paths to prove termination and correction (due to the presence of negative prices). In order to push our algorithm as far as we can, we introduce the class of negative-reset-acyclic games for which we obtain the same result: as a particular case, we can solve all priced timed games with one clock for which the clock is reset in every cycle of the underlying region automaton. As future works, it is appealing to solve the full class of priced timed games with arbitrary weights and one clock. We have shown why our technique seems to break in this more general setting, thus it could be interesting to study the difficult negative cycles without reset as their own, with different techniques.
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[^1]:    ${ }^{2}$ Here we differ from [BLMR06] where $L_{u} \subseteq L_{\text {Max }}$.
    ${ }^{3}$ Not that in our setting, an affine function is of the form $f(\nu)=a \times \nu+b$.
    ${ }^{4}$ Observe that this last restriction is not without loss of generality in the case of PTGs. While all timed automata $\mathcal{A}$ can be turned into an equivalent (with respect to reachability properties) $\mathcal{A}^{\prime}$ whose clocks are bounded $\left[\mathrm{BFH}^{+} 01\right]$, this technique cannot be applied to PTGs, in particular with arbitrary prices.
    ${ }^{5}$ Throughout the paper, we often drop the $\mathcal{G}$ in the subscript of several notations when the game is clear from the context.

[^2]:    ${ }^{6}$ For this result, the order does not depend on the owner of the location, but on the fact that $\ell^{\star}$ has minimal price amongst locations of $\mathcal{G}$.

