# The Inefficiency of Nash and Subgame Perfect Equilibria for Network Routing 

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#### Abstract

This paper provides new bounds on the quality of equilibria in finite congestion games with affine cost functions, specifically for atomic network routing games. It is well known that the price of anarchy equals exactly $5 / 2$ in general. For symmetric network routing games, it is at most $(5 n-2) /(2 n+1)$, where $n$ is the number of players. This paper answers to two open questions for congestion games. First, we show that the price of anarchy bound $(5 n-2) /(2 n+1)$ is tight for symmetric network routing games, thereby answering a decade-old open question. Second, we ask whether sequential play and subgame perfection allows to evade worst-case Nash equilibria, and thereby reduces the price of anarchy. This is motivated by positive results for congestion games with a small number of players, as well as recent results for other resource allocation problems. Our main result is the perhaps surprising proof that subgame perfect equilibria of sequential symmetric network routing games with linear cost functions can have an unbounded price of anarchy. We complete the picture by analyzing the case with two players: we show that the sequential price of anarchy equals $7 / 5$ and that computing the outcome of a subgame perfect equilibrium is NP-hard.


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Keywords: Nash and subgame perfect equilibrium • price of anarchy • congestion games • network routing

## 1. Introduction

The concept of the price of anarchy (PoA), introduced by Koutsoupias and Papadimitriou [23], has spurred a significant amount of research over the past two decades and has contributed to establishing the area algorithmic game theory. In its original version, one relates the outcome of a worst-case Nash equilibrium of a noncooperative game with selfish players to that of an optimal, centrally computable solution, using as a metric some global objective function defined on the set of possible outcomes of the game. It is well known that the lack of central coordination, translated into solutions that arise as Nash equilibria, can lead to substantial efficiency losses. This remains true even if the global objective function is the sum of the objective functions of all players.

In their highly influential work, Roughgarden and Tardos [31] prove that the price of anarchy for "nonatomic" network routing games with affine cost functions is exactly $4 / 3$. The equilibrium concept in nonatomic network routing is Wardrop's equilibrium [33], meaning that only minimum cost paths can have nonzero flow. Wardrop equilibria possess a simple characterization through variational inequalities that has been extensively used to derive price of anarchy results (e.g., Correa et al. [8]). Furthermore, they can be thought of as the limiting case of a Nash equilibrium where the total flow is routed by infinitely many players, each responsible for an infinitesimally small amount of flow. This variational inequality approach was shortly after used to pin down, to $5 / 2$, the price of anarchy of the "atomic" counterpart of the network routing problem (Awerbuch et al. [4], Christodoulou and Koutsoupias [7]).

This paper focuses on finite congestion games with affine cost functions, and specifically their special case, atomic network routing with linear cost functions. A congestion game consists of set of $n$ players and a set of $m$ resources, and each player is equipped with a feasibility system of subsets of the resources that describes the feasible choices that are available for that player. The cost of a resource is an affine, nonnegative function that
depends on the number of players that choose that resource. A solution consists of a choice of a feasible subset for each of the players, and the total cost of a player is the sum of the costs of all chosen resources. The total cost of the system is the sum of the costs of all players. In the special case of atomic network routing, the resources are arcs in a digraph, and each of the $n$ players is interested in establishing one origin-destination path. The game is symmetric if all players have the same feasibility system, and, translated into network routing, that means that all players have the same origin and destination. For general congestion games with affine cost functions, the price of anarchy is known to be equal to $5 / 2$ for $n \geq 3$ players (Awerbuch et al. [4], Christodoulou and Koutsoupias [7]), and it is at most $(5 n-2) /(2 n+1)$ if the game is symmetric (Christodoulou and Koutsoupias [7]).

This paper first closes a gap in the literature by showing that $(5 n-2) /(2 n+1)$ is indeed the exact price of anarchy for symmetric network routing games. The major part of this paper, however, is dedicated to the question what effect sequential decision making has on the quality of the resulting equilibria. Let us next elaborate on the relevance of this question.

One motivation to address the effect of sequential decision making lies in a set of recent publications, led by Paes Leme et al. [29]. They made the proposal to consider sequential versions of finite, $n$-player games and to analyze subgame perfect equilibrium (SPE) outcomes of such sequential games, instead of Nash equilibria of the simultaneous move strategic form games. Roughly speaking, subgame perfect equilibria represent the rational outcomes of a sequential game where all players act farsighted and strategically (Selten [32]). The driver behind this proposal is the insight that some of the Nash equilibria of the strategic form game cannot be realized as subgame perfect equilibria of the sequential version of the game. That leads to the notion of the sequential price of anarchy (SPoA), introduced by Paes Leme et al. [29], which could be smaller than the price of anarchy. Similar to the price of anarchy (Koutsoupias and Papadimitriou [23]), the sequential price of anarchy measures the cost of decentralization. However, whereas the price of anarchy compares the quality of a worst-case Nash equilibrium to the quality of an optimal solution, the sequential price of anarchy considers the possible outcomes of a game where players choose their strategies sequentially, in some arbitrary (albeit fixed) order. It then relates the quality of the outcome of the worst possible subgame perfect equilibrium taken over all possible orders of the players to the quality of an optimal solution. Note that for games with perfect information, which we consider here, subgame perfect equilibria can be computed via backward induction. Although this is conceptually simple, it raises an important challenge compared with the understanding of Nash or Wardrop equilibria, the main difficulty being the absence of a simple characterization such as the variational inequality mentioned earlier.

It turns out that there are interesting examples of games where sequential play and subgame perfection can lead to substantially improved guarantees on the price of anarchy and in this sense allow one to avoid what has been called the "curse of simultaneity" (Paes Leme et al. [29]). Indeed, for a handful of games, the SPoA has been proven to be lower than the PoA (de Jong and Uetz [10], de Jong et al. [11], Paes Leme et al. [29]). For some others, however, this turns out to be the opposite (Angelucci et al. [1], Bilò et al. [6]).

The motivation to analyze the sequential price of anarchy also for congestion games, and specifically for atomic network routing games, is twofold.

First, de Jong and Uetz [10] considered general congestion games with affine cost functions and showed that the sequential price of anarchy is equal to 1.5 for two players, and equal to $2+\frac{63}{488}$ for three players. That means in particular that worst-case Nash equilibria are not realizable by farsighted players in sequential play. De Jong and Uetz [10] further conjectured that the SPoA for affine congestion games is below $5 / 2$, for any number of players. Yet the only known upper bound on the SPoA for an arbitrary number of players is-trivially-the number of players, $n$.

Second, in practical situations, it often seems that sequential decision making is a more realistic model compared with simultaneous single shot games. This is certainly true for applications such as traffic and load balancing, where autonomous decisions are typically taken on top of an existing and known load of the resources, but future decisions of other players are to be expected. The model that we consider is clearly an abstraction of such concrete applications, and specifically asks what the effect is if all $n$ players act sequentially and fully farsighted. We believe it is important to understand the effect of sequential play and subgame perfection in this basic setting.

### 1.1. Our Results

As a first result, this paper answers a question that was posed, among others, by Christodoulou and Koutsoupias [7] and Bhawalkar et al. [5], and that had remained an open problem for more than a decade. Indeed, it was open whether the price of anarchy bound $5 / 2$ can also be attained (asymptotically) with symmetric network routing games, and if the upper bound $(5 n-2) /(2 n+1)$ of Christodoulou and Koutsoupias [7]
is tight. We close this gap, by deriving a lower bound equal to $(5 n-2) /(2 n+1)$ for symmetric, atomic network routing games with linear cost functions. We thereby exactly match the upper bound that was known for the case of symmetric (not necessarily network) congestion games (Christodoulou and Koutsoupias [7]). This asymptotically matches the upper bound $5 / 2$ that is known for $n \geq 3$ players for the asymmetric network routing case (Awerbuch et al. [4], Christodoulou and Koutsoupias [7]).

As to the quality of equilibria for sequentially played congestion games, we show the following. First, we pin down the exact sequential price of anarchy for the symmetric network routing case with $n=2$ players, showing that it equals $7 / 5$. This constitutes an improvement over the $3 / 2$ bound that is known for the more general, nonsymmetric case (de Jong and Uetz [10]). For a large number of players, however, and in sharp contrast to what was conjectured by de Jong and Uetz [10], we show that congestion games, and even symmetric network routing games, have an unbounded sequential price of anarchy. Indeed, we prove that even in the symmetric network routing case, that is, when all players share the same origin and destination, the sequential price of anarchy can be as large as $\Omega(\sqrt{n})$. In view of earlier results that compare Nash to subgame perfect equilibria (de Jong and Uetz [10], de Jong et al. [11], Paes Leme et al. [29]), this establishes an unexpectedly sharp separation between Nash and subgame perfect equilibria for network routing games. Subsequent recent work by Groenland and Schäfer [16] suggests that the presence of ties seems to be pivotal for the existence of such worst-case instances.

The crucial part of the proof of our main result is to come up with a "contingency plan of actions" for every player and every possible profile of actions of all previous players that indeed leads to a subgame perfect outcome. This is generally difficult, because the strategies of the players are of exponential size. We are, however, able to design a plan leading to an unbounded SPoA that can be described in a succinct manner: The core idea, that we believe may be of independent interest, is to design a "master plan" of actions that all players are supposed to follow, together with a "punishing" action that players only apply when some previous player deviates from the master plan. The main technical difficulty is to design a construction such that the punishing actions do not lead to a higher cost for the player applying it, so that subgame perfection can be achieved. To some extent, this construction may seem somewhat unrealistic from a practical viewpoint, yet the main point is to show that such a construction is possible at all for atomic congestion games.

We finally also address the computational complexity of subgame perfection, showing that the computation of the outcome of a subgame perfect equilibrium is an NP-hard problem. Although we know from Paes Leme et al. [29] that computing subgame perfect equilibria can be PSPACE-hard in general congestion games, that reduction requires a nonconstant number of players. Our result shows that the problem remains at least NP-hard for symmetric network routing games, and even when the number of players is only two. This contrasts with the fact that a pure Nash equilibrium can be efficiently computed in these games through a minimum cost flow computation (Fabrikant et al. [12]). Interestingly, our analysis also shows that outcomes of subgame perfect equilibria are generally not Nash equilibria of the simultaneous single shot game, as opposed to, for example, crowding games, as studied by Milchtaich [26]. Conditions under which this is actually the case for network routing games have been given by Groenland and Schäfer [16].

### 1.2. Outline

We first discuss in Section 2 the related literature with respect to the sequential price of anarchy. We then formally introduce the model along with some necessary preliminary definitions and insights in Section 3. Section 4 addresses the (nonsequential) atomic network routing game and shows that the regular price of anarchy equals $(5 n-2) /(2 n+1)$ when the game is symmetric. In Section 5 , we start discussing the sequential network routing game for the case when there are only $n=2$ players. We prove that the sequential price of anarchy is precisely $7 / 5$ and show that computing an outcome from a subgame perfect equilibrium is NP-hard. In Section 6, we prove the main result of this paper, namely, that the sequential price of anarchy of symmetric network routing games with linear cost functions is generally unbounded. In Section 7, we conclude with some remarks and open problems.

## 2. Related Work on the Sequential Price of Anarchy

Network routing games and congestion games form a cornerstone class of problems in the field of algorithmic game theory. Many variants and extensions on these games have been proposed, and the body of literature on these games is so large that it is impossible to survey it or summarize its most important articles in this section. We refer to the recent survey by Fotakis [14], which discusses various results on the price of anarchy and the price of stability, including more general variants such as weighted congestion games, nonlinear cost functions, etc. In the remainder of this section, we focus on recent work on sequential games and subgame perfect equilibria, as the present paper falls most naturally into this line of research.

The idea to consider subgame perfect equilibria of sequential games as an alternative to rule out some potentially "unrealistic" Nash equilibria was initiated by Paes Leme et al. [29]. They introduced the notion of the sequential price of anarchy (abbreviated SPoA) and proved it to be relevant for three classes of games: In consensus games, players choose one of two alternatives, and gain utility by picking the same alternative as other players. The authors show that consensus games have an SPoA of 1 as opposed to an unbounded PoA. In machine cost sharing games with fair cost allocation, each player chooses a set of machines, and players distribute the cost of each machine equally. The authors show that the SPoA is $\Theta(\log n)$, an improvement over the PoA, which is equal to $n$ for these games (where $n$ is the number of players). In unrelated machine scheduling games, each player controls a job and needs to choose a machine to run the job on. Processing times of the jobs are machine dependent, and jobs are interested in the total load of the selected machine. Unrelated machine scheduling games have an unbounded PoA, but the SPoA lies in between $\Omega(\sqrt{n})$ and $m \cdot 2^{n}$, where $m$ is the number of machines. The paper (Paes Leme et al. [29]) also contains the result that subgame perfect equilibria are PSPACE-hard to compute in general.

Later, Bilò et al. [6] showed that the improvements obtained by Paes Leme et al. [29] should be relativized, as they hold only for the class of generic games. ${ }^{1}$ Specifically, in the nongeneric case, consensus games as well as the related class of cut games have an SPoA of 3, whereas machine cost sharing games with fair cost allocation have an SPoA between $\Omega\left(n+1-H_{n}\right)$ and $n$, and unrelated machine scheduling games have an SPoA between $2^{\Omega(\sqrt{n})}$ and $2^{n}$. Note that the latter upper bound is also an improvement for the generic case.

Closest to our work is the study of congestion games with few players by de Jong and Uetz [10]. That paper triggered the question on the actual value of the price of anarchy for atomic congestion games that is ultimately answered here. The authors show that for two players, the SPoA equals $3 / 2$, and for three players, it equals $2+63 / 488 \approx 2.13$. Hence, in those cases, the SPoA is less than the PoA, which equals 2 and $5 / 2$, respectively (Christodoulou and Koutsoupias [7]). It was conjectured by de Jong and Uetz [10] that the PoA for affine congestion games is below $5 / 2$, but Kolev [22] showed that when the number of players is four or more, the SPoA already exceeds 2.5 .

We conclude this introductory section with a brief account of models for which the effect of sequential play on the quality or existence of equilibria has been analyzed:

- A set of authors that first addressed the effect of sequential decisions in network routing games are Olver [27], Harks et al. [19], Harks and Vegh [18], as well as Farzad et al. [13]. Without going into too much detail, their models differ in that the cost of a given player depends on preceding players but not on future players, just like in traffic situations where players enter streets in sequence, and a player only influences players that are behind.
- Hassin and Yovel [20] consider machine scheduling and focus on the special case of related machines. The authors prove that for this model, the SPoA is at most 2, whereas the PoA is $\Theta(\ln n / \ln \ln n)$. Moreover, the SPoA decreases to $4 / 3$ when each machine always executes the shortest jobs first. Their analysis relies on well-known approximation results for list scheduling by Graham [15].
- The effect of sequential play was also analyzed for a throughput scheduling model by de Jong et al. [11]. There, each player owns a set of machines, and players select jobs with individual values that must be scheduled in given time windows. The authors show that for identical machine scheduling, the SPoA is $e /(e-1)$, an improvement over the PoA, which is 2.
- Avni et al. [3] study another variation of congestion games. The difference is that in their model, players make multiple, incremental decisions, and each time choose a single resource that extends the current solution. These decisions are in scheduled phases, and within each phase, the players make decisions in $k \leq n$ subsequent turns, where at least one player makes a choice in each turn. In some sense, this model mixes sequential and simultaneous decision making, and the main results of Avni et al. [3] are on the existence and nonexistence of (pure strategy) subgame perfect equilibria.
- Rahn and Schäfer [30] study tree graph coordination games, where players are nodes in a graph. Each player chooses a color, and obtains utility equal to the number of neighbors sharing her color. The authors prove that the SPoA is 2 , whereas the PoA is unbounded.
- Yet another context is that of item bidding games. Here players bid on a set of public projects, aiming to have their personal preferred project selected. Lucier et al. [25] prove, among other things, that even though the PoA can be as large as $n-1$, any subgame perfect equilibrium in the sequential version is optimal, yielding an SPoA of 1 .
- Isolation games were studied by Angelucci et al. [1]. These are games in which players have to choose a point in a metric space and are interested in maximizing the distance to any other player. The authors show that, depending on the definition of the players' utility functions (sum or minimum distance) and social cost (utilitarian or egalitarian), the SPoA is higher than the PoA in some cases and lower in other cases.
- Variants of sequential matching games were studied by Haeringer and Wooders [17] and in the very recent papers by Kawase et al. [21]. Haeringer and Wooders [17] characterize Nash and subgame perfect equilibria in a game where firms sequentially propose workers to be assigned to them, after which a second stage starts in which the workers make their acceptance and rejection decisions. Kawase et al. [21] study the problem to compute a subgame perfect equilibrium in a similar type of game where the difference is that a worker immediately makes his decision after each offer. They show various tractability results and PSPACE-hardness results.
- Subsequent to this paper, recent works by Groenland and Schäfer [16] suggest bridging the gap between far-sighted behaviour and myopic greedy best-response behaviour. They assume that for some parameter $k$, each player chooses a best response with respect to the action profile that arises when the $k$ subsequent players play subgame perfectly. The authors establish conditions under which the action profiles resulting from such behaviour correspond to Nash equilibria, thereby allowing them to import known results for the price of anarchy. In particular, for a special class of graphs, their work shows that the absence of ties guarantees that subgame perfect equilibria are also Nash equilibria, and hence have a bounded sequential price of anarchy.


## 3. Model and Notation

Throughout this paper, we consider a special class of finite, atomic congestion games, namely, symmetric atomic network routing games with linear cost functions. The input of an instance $I \in \mathscr{I}$ consists of a directed graph $G=(V, E)$, with designated source and target nodes $s, t \in V$, and for each arc $e \in E$ a linear cost (or latency) function $c_{e}(y)=a_{e} y$ with linear cost coefficient $a_{e}$; that is, the cost of an arc equals $a_{e}$ times the number of players using it. There are $n$ players that all want to travel from the same origin $s$ to destination $t$, so that the possible actions of all players consist of all directed ( $s, t$ ) paths in $G$. Hence, all players have the same set of actions and the game is symmetric. We will denote by $m$ the number of arcs $|E|$. We refer to the path that a player $i$ chooses the action $A_{i}$ of player $i$, and $A=\left(A_{1}, \ldots, A_{n}\right)$ is a vector with one action for each player, also called an outcome or action profile.

The cost of a player $i$ for choosing a specific $(s, t)$ path $A_{i}$ depends on the number of players on each arc on that path. Specifically, for an outcome $A=\left(A_{1}, \ldots, A_{n}\right)$, let

$$
n_{e}(A):=\sum_{i=1}^{n}\left|A_{i} \cap\{e\}\right|
$$

denote the number of players using arc $e$; then the cost of that arc for each player using it equals $a_{e} n_{e}(A)$, and therefore the cost for player $i$, choosing path $A_{i}$, is defined as

$$
c_{i}(A)=\sum_{e \in A_{i}} a_{e} \cdot n_{e}(A) .
$$

For the sake of strengthening some of our results, sometimes we assume that the cost functions are affine instead of linear. ${ }^{2}$ Affine cost functions are of the form

$$
c_{i}(A)=\sum_{e \in A_{i}}\left(a_{e} \cdot n_{e}(A)+b_{e}\right),
$$

where $b_{e} \geq 0$ can be thought of as a player-independent activation cost associated to an arc $e \in E$.
We define the social cost of an action profile $A$ as $C(A)=\sum_{i=1}^{n} c_{i}(A)$, that is, the sum of the costs of the players. In other words, we consider a utilitarian social cost function. This is a standard model, yet note that it differs from the egalitarian makespan objective as studied, for example, by Koutsoupias and Papadimitriou [23].

A pure Nash equilibrium is an outcome $A$ in which no player can decrease her costs by unilaterally deviating, that is, switching to an action that is different from $A_{i}$. The price of anarchy (PoA) (Koutsoupias and Papadimitriou [23]) measures the quality of any Nash equilibrium relative to the quality of a globally optimal allocation, OPT. Here OPT is an outcome minimizing $C(\cdot)$. More specifically, for an instance $I$,

$$
\begin{equation*}
\operatorname{PoA}(I)=\max _{N E \in N E(I)} \frac{C(N E)}{C(O P T)}, \tag{1}
\end{equation*}
$$

where $N E(I)$ denotes the set of all Nash equilibria for instance $I$. The price of anarchy of a class of instances $\mathscr{I}$ is defined by $\operatorname{PoA}(\mathscr{F})=\sup _{I \in \mathscr{F}} \operatorname{PoA}(I)$.

In this paper, our goal is to evaluate the quality of subgame perfect equilibria of an induced extensive form game that we call the sequential version of the game (Kuhn [24], Selten [32]). In the sequential game, players choose an action from the set of $(s, t)$ paths, but instead of doing so simultaneously, they choose their actions in
an arbitrary predefined order $1,2, \ldots, n$, so that the $i$ th player must choose action $A_{i}$, observing the actions of players preceding $i$, but of course not knowing the actions of the players succeeding her. However, because players are fully rational and fully informed, at equilibrium they anticipate the successors' behaviour and therefore make optimal strategic choices by fully anticipating all followers' actions.

A strategy $S_{i}$ then specifies for player $i$ the full contingency plan of actions she would choose for each potential choice of actions $A_{<i}:=\left(A_{1}, \ldots, A_{i-1}\right)$ chosen by her predecessors. We use $S_{i}\left(A_{<i}\right)$ to denote the action that $i$ plays under strategy $S_{i}$ when $A_{<i}$ is the vector of actions chosen by players $1, \ldots, i-1$. We refer to a choice of strategies $S=\left(S_{1}, \ldots, S_{n}\right)$ by each of the players as a strategy profile. Note the explicit distinction between action (profile) and strategy (profile). The outcome resulting from $S$ is then the set of actions chosen by the players when they play according to the strategy profile $S$.

Subgame perfect equilibria, defined by Selten [32], are defined as strategy profiles $S$ that induce Nash equilibria in any subgame of the extensive form game. In other words, a strategy profile $S$ is a subgame perfect equilibrium if for all $i$ and for any choice of actions $A_{<i}$ of players $1, \ldots, i-1$, player $i$ cannot decrease her cost by switching to an action different from $S_{i}\left(A_{<i}\right)$, in the subgame where the actions of $1, \ldots, i-1$ are fixed to $A_{<i}$, and players $i+1, \ldots, n$ play strategies $\left(S_{i+1}, \ldots, S_{n}\right)$.

Subgame perfect equilibria reflect farsighted strategic behaviour of players that observe the state of the game and reason strategically about choices of subsequent players, always choosing the action that will minimize their individual cost. Analogous to (1), the sequential price of anarchy of an instance $I$ is defined by

$$
\begin{equation*}
\operatorname{SPoA}(\mathrm{I})=\max _{S P E \in S P E(I)} \frac{\mathrm{C}(\mathrm{SPE})}{\mathrm{C}(\mathrm{OPT})}, \tag{2}
\end{equation*}
$$

where $\operatorname{SPE}(I)$ denotes the set of all outcomes of subgame perfect equilibria of instance $I$. The sequential price of anarchy of a class of instances $\mathscr{I}$ is defined as by Paes Leme et al. [29] by $\operatorname{SPoA}(\mathscr{F})=\sup _{\mathrm{I} \in \mathscr{\mathscr { F }}} \operatorname{SPoA}(I)$. Throughout this paper, when the class of instances is clear from the context, we write PoA and SPoA.

Extensive form games can be represented in a game tree (see Figure 1 for an example), with the nodes on level $i$ representing the possible states of the game that a single player $i$ can encounter, and the arcs emanating from any node representing the possible actions of that player in the given state. The nodes of the game tree are called information sets or states. We will refer to a state by a pair $\left(A_{<i}, i\right)$, where $A_{<i}$ is the choice of actions of the players $1, \ldots, i-1$ in that state, and $i$ is the next player who has to choose her action. Because we deal with a game with perfect information, subgame perfect equilibria can be computed by backward induction. In particular, it is known that subgame perfect equilibria always exist; see, for example, Osborne [28]. Note, however, that if $S$ is a subgame perfect equilibrium, the resulting outcome $A$ need not be a Nash equilibrium of the corresponding strategic form game, as will also be discussed in Section 5.

## 4. The Price of Anarchy for Symmetric Network Routing

In this section, we first focus on the regular price of anarchy PoA as defined in (1) for symmetric network routing games with linear cost functions. We show that it equals $(5 n-2) /(2 n+1), n$ being the number of players. This resolves an open question regarding the price of anarchy of network congestion games of

Figure 1. Game tree for a symmetric sequential two player game. Nodes are the states. Note that $A_{11}$ and $A_{21}$ are actions of players 1 and 2, respectively, but denote the same action (recall that we have a symmetric game). The same holds for $A_{12}$ and $A_{22}$. Fat lines denote a subgame perfect strategy $S=\left(S_{1}, S_{2}\right)$, where $S_{1}=A_{12}, S_{2}\left(A_{11}\right)=A_{21}$, and $S_{2}\left(A_{12}\right)=A_{22}$. The outcome resulting from $S$ is $\left(A_{12}, A_{22}\right)$, the rightmost path of the game tree.


Christodoulou and Koutsoupias [7] (see also Bhawalkar et al. [5]). Indeed, it is well known that the price of anarchy of the more general class of linear symmetric (not necessarily network) congestion games is at most $(5 n-2) /(2 n+1)$ (Christodoulou and Koutsoupias [7]). It has not been clear, however, whether this bound can also be attained with symmetric linear network routing games. Here we settle that issue and give a matching lower bound of $(5 n-2) /(2 n+1)$. Surprisingly, the lower bound is conceptually simpler than the one previously provided for the more general class of (nonnetwork) affine congestion games (Christodoulou and Koutsoupias [7]).
Theorem 1. The price of anarchy of symmetric linear network routing games with $n$ players equals $(5 n-2) /(2 n+1)$.
Proof. We construct the following family of instances. For three players, the instance (along with the optimal and equilibrium strategies) is depicted in Figure 2. In general, let $n$ be the number of players, and consider an instance in which there are $n$ disjoint paths from the source $s$ to the sink $t$, which we refer to as principal paths. These paths are all composed of $2 n-1$ arcs and thus $2 n$ nodes ( $s$ being the first and $t$ being the last), so we denote by $e_{i j}$ the $j$ th arc of the $i$ th path, for $i=1, \ldots, n$ and $j=1, \ldots, 2 n-1$, and by $v_{i, j}$ the $j$ th node of the $i$ th path, for $i=1, \ldots, n$ and $j=1, \ldots, 2 n$. There are $n \cdot(n-1)$ additional connecting arcs that connect these paths: there is an arc from $v_{i, 2 k+1}$ to $v_{i-1,2 k}$ for $k=1, \ldots, n-1$, where $i-1$ is taken $\bmod (n)$. This defines the network. The cost coefficients on the arcs are set as follows. Arcs $e_{i, 1}$ (that start from $s$ ) have cost coefficient $a_{e}=2$, arcs $e_{i, 2 n-1}$ (that end in $t$ ) have cost coefficient $a_{e}=2$, and arcs $e_{i, j}$ with $1<j<2 n-1$ have cost coefficient $a_{e}=1$. All connecting arcs have cost coefficient $a_{e}=0$.

It is easy to check that the optimal solution in this instance is to route one player in each of the principal paths. Because in this solution no two players intersect in any arc, its cost can be computed as $n(2+2 n-$ $3+2)=n(2 n+1)$. On the other hand, a Nash equilibrium is obtained when each player $k$ follows the following path: she starts with arcs $e_{k, 1}$ and $e_{k, 2}$; then uses all arcs of the form $e_{k+j, 2 j}, e_{k+j, 2 j+1}$, and $e_{k+j, 2 j+2}$ for $j=1, \ldots, n-2$; and finishes with arcs $e_{k+n-1,2 n-2}$ and $e_{k+n-1,2 n-1}$ and uses all the required connecting arcs. Here, the additions on the principal paths index are taken $\bmod (n)$. In this solution, every arc $e_{i, j}$ with $j$ even is used exactly twice, whereas every other arc is used only once. Thus, the social cost is $n(n-1) \cdot 4+n(n-2) \cdot 1+2 n \cdot 2=n(5 n-2)$. It immediately follows that the PoA of the instance is at least $(5 n-2) /(2 n+1)$.

The remainder of the proof consists in checking that the latter path choices indeed result in a Nash equilibrium. Consider a player, which by symmetry we may assume is player 1, and let us evaluate possible deviations from the current path, which we call the zigzag path $Z$. Note that in any path an arc $e_{i, 2 j-1}$ is always followed by $e_{i, 2 j}$ for $2 \leq j \leq n-1$, and that player 1 evaluates the joint cost of these two arcs as either 5 (if not in the zigzag path) or 3 (if in the zigzag path). Now assume player 1 follows a path $P$ not intersecting $Z$ in arcs of
Figure 2. A lower bound instance for the PoA for $n=3$. Players travel from $s$ to $t$.

the form $e_{i, 2 j-1}$ for $j=1, \ldots, n$; then the cost of this path is at least $(n-2) \cdot 5+8=5 n-2$ (the extra 8 comes from the arc starting in $s$ and the arc ending in $t$ ), and thus the deviation is not profitable. Therefore, we may assume that $P$ and $Z$ do intersect in arcs of the form $e_{i, 2 j-1}$. Because these arcs are in $Z$, they actually are of the form $e_{i, 2 i-1}$. Assume there are two such arcs, and consider two of these intersection arcs $e_{i, 2 i-1}$ and $e_{k, 2 k-1}$ such that there is no other intersection arc on the path segment of $P$ that runs from $e_{i, 2 i-1}$ to $e_{k, 2 k-1}$. The cost of the restricted $Z$ path between nodes $v_{i, 2 i}$ and $v_{k, 2 k-1}$ is $3+5(k-i-1)$, whereas path $P$ has to cost at least $5(k-i-1)$ just to get to a node of the form $v_{l, 2 k-1}$ plus what it needs to pay to get to the principal path $k$, which is at least 3. The total cost is thus at least $3+5(k-i-1)$, implying that the deviation $P$ restricted to the segment between any two edges intersecting with $Z$ is not profitable. Finally, we consider the subpath between $s$ and the first such intersection, say, $e_{i, 2 i-1}$ (and symmetrically between the last and $t$ ). In this case, the cost of the restricted $Z$ path between nodes $s$ and $v_{i, 2 i}$ is $6+(i-2) 5$, whereas the cost of $P$ is at least $4+3+(i-2) 5$. Here, the 4 comes from the first arc and the 3 from the second arc. Again, the deviation cannot be profitable. We thus conclude that $Z$ is indeed a best response, and thus we have a Nash equilibrium.

## 5. Sequential Games and Equilibria for $n=2$ Players

As a way to illustrate the difficulties in arguing about subgame perfect equilibria in general, we first focus on the special case with only two players. Doing that, we point out two phenomena that clarify the fundamental difference between the concept of subgame perfect equilibrium and that of Nash equilibrium.

First, we derive a simple instance in which the resulting actions of a subgame perfect equilibrium do not correspond to a Nash equilibrium. This is in contrast to the case of parallel links (de Jong and Uetz [10]) and also to the crowding games as studied by Milchtaich [26]. This effect is the major complication in analyzing subgame perfect equilibria, and it seems it can be circumvented if the underlying game is generic, meaning that no player is ever confronted with a tie; see Milchtaich [26], Bilò et al. [6], and Groenland and Schäfer [16] for network routing games in particular.

Based on this particular instance, we additionally prove that the sequential price of anarchy for the two player case equals $7 / 5$. This is smaller than the regular price of anarchy, which is equal to $8 / 5$ [7], and also smaller than the sequential price of anarchy for the asymmetric network routing case, which equals $3 / 2$ (de Jong and Uetz [10]). ${ }^{3}$ Second, we show that even in the two-player case, computing the outcome of a subgame perfect equilibrium is NP-hard.

### 5.1. The Sequential Price of Anarchy

Consider the two-player instance depicted in Figure 3, with five vertices and eight arcs. The vertices 1, 2, 3, 4, 5 are numbered from left to right and from top to bottom so that $s=1$ and $t=5$. The linear cost coefficients $a_{e}$ are given by the numbers next to the respective arcs $e$.

Figure 3. Lower bound example for two players. Each arc is labeled with the coefficients of the linear cost function associated with it. There are two parallel arcs emanating from $s$ : The purpose of the arc with the higher coefficient is to cause player 2 to be indifferent between the two arcs in case player 1 uses the arc with the lower coefficient. This ensures that there exists an SPE that results in the action profile depicted on the right-hand side.


SPE


There are two parallel arcs between vertices 1 and 2 . In the discussion below, whenever we specify that a player takes a path that includes an arc from 1 to 2 , that player uses the parallel arc with the lower of the two coefficients. It can be easily verified that the following is a subgame perfect equilibrium:

- Player 1 chooses path ( $1,2,3,4,5$ ).
- Player 2 chooses
- $(1,5)$ if player 1 chooses $(1,2,3,4,5)$,
- $(1,2,4,5)$ if player 1 chooses $(1,2,3,5)$,
$-(1,2,3,5)$ if player 1 chooses ( $1,2,4,5$ ),
-any (best-response) path for all remaining choices of player 1.
In this equilibrium outcome, player 1 chooses the dashed path on the right, that is, path ( $1,2,3,4,5$ ), whereas player 2 chooses the dotted path on the right, which is simply the arc going directly from 1 to 5 . Interestingly, one may think that player 1 has an incentive to deviate to the path $(1,2,3,5)$ because the cost of going straight from 3 to 5 is 0 . However, if player 1 does this, player 2 would pick path ( $1,2,4,5$ ), and therefore player 1 's cost would still be 3 . This implies that indeed the outcome of the subgame perfect equilibrium is not a Nash equilibrium. Note furthermore that player 1's cost is 3 and player 2's is 4, for a total social cost of 7 , whereas in the socially optimal situation, depicted on the left of Figure 3, the social cost is 5. So in particular this instance shows that the SPoA is at least $7 / 5$.

In the above instance, the subgame perfect equilibrium is not unique. However, the latencies can be slightly perturbed so that uniqueness is achieved, while the cost of the equilibrium remains arbitrarily close to 7 and that of the optimum remains arbitrarily close to 5 . To this end consider the same instance but change the cost coefficient of arc $(1,2)$ from 2 to $2+\epsilon$, that of arc $(1,5)$ from 4 to $4+\epsilon$, and those of arcs $(2,4)$ and $(3,5)$ from 0 to $\epsilon$.

With the latter observation, not only the sequential price of anarchy but also the sequential price of stability ${ }^{4}$ equals $7 / 5$ in the two-player case. This is because it is possible to prove a matching upper bound, even for the more general class of symmetric affine (not necessarily network) congestion games. It uses a proof technique based on linear programming, but is nonetheless different from the proof used by de Jong and Uetz [10], where linear programming was also used to derive upper bounds on the SPoA. The main result of this section is the following.
Theorem 2. The sequential price of anarchy of symmetric linear network routing games and symmetric affine congestion games is $7 / 5$ when there are two players only.
Proof. It follows from the example above that the SPoA for two-player linear network routing games is at least 7/5, so it remains only to show that $7 / 5$ is also an upper bound. The proof goes by establishing a set of valid inequalities that must be fulfilled for any outcome of a subgame perfect equilibrium $\left(A_{1}, A_{2}\right)$ and any outcome $\left(A_{1}^{*}, A_{2}^{*}\right)$ that minimizes the total cost. After applying several combinatorial arguments, we arrive at the matching upper bound of $7 / 5$ via linear programming. The details of these arguments that complete the proof of Theorem 2 , together with a combinatorial argument to show that the underlying linear program has $7 / 5$ as an optimal solution, can be found in Lemma A. 1 in the appendix.

### 5.2. Hardness of Computing Subgame Perfect Equilibria

Notice that the encoding of subgame perfect strategies can, in general, require superpolynomial space in terms of the input size of a network routing game. This is even the case for two players, for example, if the first player has a superpolynomial number of possible actions, that is, $(s, t)$ paths. Then, for each of these potential actions of player 1, a subgame perfect equilibrium needs to prescribe the respective actions taken by player 2. We head for a meaningful statement, however, with respect to the input size of a network routing game, and not the output. Therefore, we consider the computational problem to only output the outcome resulting from a subgame perfect equilibrium. This exactly corresponds to a single path in the game tree, which for two players has depth two. This outcome has polynomial size, as it consists of just one path in the network $G$ per player. The problem to compute such an equilibrium path in the game tree, however, turns out to be hard.
Theorem 3. Computing an action profile resulting from a subgame perfect equilibrium is (strongly) NP-hard for any number of players $n \geq 2$.
Proof. We prove the theorem by a reduction from the Hamiltonian path problem, where the task is to determine whether a given undirected graph has a Hamiltonian path, that is, a path that visits each vertex of the graph exactly once. Consider any instance of Hamiltonian path on undirected graph $G=(V, E)$, where we assume without loss of generality (w.l.o.g.) that $G$ is connected and has at least one edge. We construct the following game: There are $n$
players. There are two copies $v^{\prime}, v^{\prime \prime}$ for each node $v \in V$. There is also a source node $s$, a sink node $t$, and a node $s^{\prime}$. We define $m=2|V|+1$, and $\epsilon=1 /(|E|-1)$. There is an arc with cost coefficient $m$ from $s$ to $s^{\prime}$, and an arc with cost coefficient $(2 m+3) /(n-1)$ from $s$ to $t$. For each $v \in V$, there is an arc with cost coefficient 0 from $s^{\prime}$ to $v^{\prime}$, an arc with cost coefficient 2 from $v^{\prime}$ to $v^{\prime \prime}$, and an arc with cost coefficient 0 from $v^{\prime \prime}$ to $t$. Moreover, for each arc $(u, v) \in E$, there is an arc with cost coefficient $\epsilon$ from $u^{\prime \prime}$ to $v^{\prime}$. This reduction is shown in Figure 4.

We claim that in an outcome resulting from a subgame perfect equilibrium, player 1 chooses all arcs ( $v^{\prime}, v^{\prime \prime}$ ) that correspond to all $v \in V$. If the graph is Hamiltonian, she will choose these arcs exactly in the order of a Hamiltonian path, and otherwise will have to traverse at least one arc $\left(v^{\prime}, v^{\prime \prime}\right)$ twice. Moreover, all subsequent players choose the arc $(s, t)$.

Let us argue that, indeed, this is the outcome of a subgame perfect equilibrium. First note that if for at least one node $v$, player 1 does not choose arc $\left(v^{\prime}, v^{\prime \prime}\right)$, then there is some successor that will choose the path $\left(s, s^{\prime}, v^{\prime}, v^{\prime \prime}, t\right)$. Let us call this player $j$. Then $j$ has a cost of at most $2 m+2$, because all other players will play $(s, t)$. The latter is due to the high cost of $3 m$ that the third player would have if she were to choose arc ( $s, s^{\prime}$ ) (whereas choosing $(s, t)$ would guarantee her a lower cost of at most $2 m+3$ ). So, in this case, player 1 has a cost of at least $2 m+2$.

On the other hand, if player 1 chooses all arcs $\left(v^{\prime}, v^{\prime \prime}\right)$, then for any succeeding player, choosing any path using an arc $\left(v^{\prime}, v^{\prime \prime}\right)$ would yield her a cost of at least $2 m+4$. In that case, choosing ( $\left.s, t\right)$ is always a better option for any succeeding player, because doing so guarantees that her cost is at most $2 m+3$.

Now, suppose there exists a Hamiltonian path in G. Let us look at the cost of player 1 if she chooses the arcs $\left(v^{\prime}, v^{\prime \prime}\right)$ that correspond to all $v \in V$, in the order of the Hamiltonian path. Then player 1's cost is at most $m+2|V|+\epsilon|V-1|<2 m$, because, as we showed, succeeding players will choose $(s, t)$.

If no Hamiltonian path exists, then for player 1, any path that contains all arcs ( $\left.v^{\prime}, v^{\prime \prime}\right)$ uses at least $|V|$ arcs with cost coefficient $\epsilon$, because she will have to use at least one arc $\left(v^{\prime}, v^{\prime \prime}\right)$ twice. This yields player 1 a cost of at least $m+2|V|+\epsilon|V|$.

From the above we conclude the following: if there is a Hamiltonian path in $G$, then there is an action $A_{1}$ for player 1 that gives her a cost of $m+2|V|+\epsilon|V-1|$ when the succeeding players all play a subgame perfect strategy profile. Moreover, playing any action other than $A_{1}$ will give player 1 a strictly higher cost when the succeeding players all play a subgame perfect strategy profile. Therefore, if there exists a Hamiltonian path, player 1 plays $A_{1}$ in a subgame perfect equilibrium and has a cost of $m+2|V|+\epsilon|V-1|$. If there is no Hamiltonian path in $G$, then in any subgame perfect equilibrium, player 1 has a cost of at least $m+2|V|+\epsilon|V|$.

Hence, if we were able to compute the outcome of a subgame perfect equilibrium in polynomial time, we could verify in polynomial time whether the cost of player 1 equals $m+2|V|+\epsilon|V-1|$ or not. This would allow us to decide the Hamiltonian path problem in polynomial time.
Recall that membership in NP is not clear in general, because it is generally not clear how to verify subgame perfectness in polynomial time. However, if we define a decision problem SPE-DEC that asks whether the first

Figure 4. Reducing a Hamiltonian path instance to an $n$-player network routing game.

player can guarantee herself a cost at most equal to some threshold $k$ in a subgame perfect equilibrium, we can also show the following.

Theorem 4. Decision problem SPE-DEC is NP-complete for the case of two players.
Proof. By Theorem 3, SPE-DEC is NP-hard, so we need to prove only that SPE-DEC is contained in NP. We use the subgame perfect action profile $\left(A_{1}, A_{2}\right)$ as a certificate for a yes instance. We can verify in polynomial time, by a shortest-path algorithm, that $A_{2}$, the action chosen by player 2, is subgame perfect. For player 1, we do not have a way to verify that $A_{1}$ is a subgame perfect action, but we do not need to: we simply verify if the cost of player 1 is indeed at most $k$ for the action profile $\left(A_{1}, A_{2}\right)$. If yes, player 1 can guarantee herself cost at most $k$ by choosing $A_{1}$, so in any subgame perfect equilibrium, player 1 will have cost at most that much. Hence, we can verify the validity of the certificate for any yes instance in polynomial time.

## 6. The Sequential Price of Anarchy for Symmetric Network Routing Games

We now prove our main result, namely, that the SPoA of symmetric network routing games is unbounded.
Theorem 5. The sequential price of anarchy of symmetric network routing games with linear cost functions is not bounded by any constant.

We prove the theorem by constructing a sequence of lower bound instances where the sequential price of anarchy gets arbitrarily large. Intuitively, the construction of these instances works as follows. There are slightly more players than disjoint $(s, t)$ paths. As an effect of our construction, it will turn out that the last player has to necessarily share every arc in her chosen action with one other player. That will result in the situation that this player can credibly "threaten" any other player $j$ by choosing the arcs that player $j$ chooses, if player $j$ does not stick to a certain action. More generally, we extend this idea so that a whole group of players can force a common predecessor into a certain action. This is achieved in such a way that the "concerted" threatening is not too expensive for every single threatener, but very expensive for the common predecessor. Altogether, the goal of the construction is to incentivize a large number of players to choose a set of arcs, much larger than in the optimal outcome, so as to drive the sequential price of anarchy to infinity. The challenging aspect is to make sure that this is indeed a subgame perfect equilibrium.

### 6.1. Definition of Instance $\Gamma_{X}$

Formally, to obtain a sequential price of anarchy of $x$, where $x \geq 4$ is a square number, we construct the following instance $\Gamma_{x}$ : Let $p$ be a sufficiently large integer. There are $n=p \sqrt{x}+5 x^{2}$ players. The network consists of $x$ segments $R_{i}, i \in[x]$. Segment $R_{i}$ consists of $2\left(1+p \sqrt{x}+4 x^{2}\right)$ nodes $\left\{i,(2 i, 1),(2 i, 2), \ldots,\left(2 i, p \sqrt{x}+4 x^{2}\right),(2 i+\right.$ $\left.1,1),(2 i+1,2), \ldots,\left(2 i+1, p \sqrt{x}+4 x^{2}\right), i+1\right\}$. Note that node $i+1$ is in both segments $R_{i}$ and $R_{i+1}$. The arc set is defined as follows:

- There is an arc $e$ with cost coefficient $a_{e}=0$ from node $i$ to node ( $2 i, j$ ) for all $j \in\left\{1, \ldots,\left(p \sqrt{x}+4 x^{2}\right)\right\}$.
- There is an arc $e$ with cost coefficient $a_{e}=1 / x$, from $(2 i, j)$ to $(2 i+1, j)$ for all $j \in\left\{1, \ldots,\left(p \sqrt{x}+4 x^{2}\right)\right\}$.
- There is an arc $e$ with cost coefficient $a_{e}=0$ from $(2 i+1, j)$ to $i+1$ for all $j \in\left\{1, \ldots,\left(p \sqrt{x}+4 x^{2}\right)\right\}$.
- There is an arc $e$ with cost coefficient $a_{e}=0$ from $(2 i+1, j)$ to $(2 i, k)$ for all $j \in\left\{1, \ldots,\left(p \sqrt{x}+4 x^{2}\right)\right\}$ and for all $k \in\left\{j+1, \ldots,\left(p \sqrt{x}+4 x^{2}\right)\right\}$.

Note that between any two nodes $i, i+1$, there exist $2^{p \sqrt{x}+4 x^{2}}$ different paths: one for every subset of arcs with cost coefficient $1 / x$ of segment $R_{i}$. For brevity, from now on, when we refer to arcs, we mean the arcs of which the cost coefficient is not identically zero, that is, arcs with cost coefficient $1 / x$.

Node 1 is the source $s$, and node $x+1$ is the sink $t$. Now any feasible action of a player consists of at least one arc from each segment $R_{i}, i \in[x]$. This example is shown in Figure 5.

In the remainder of the section, we say that in a state $\left(A_{<i}, i\right)$, an arc $e$ is free if no player in $[i-1]$ has chosen $e$ in her action, that is, there does not exist an $i^{\prime} \in[i-1]$ such that $e \in A_{i^{\prime}}$.

### 6.2. Optimal Social Cost of $\Gamma_{x}$

In the optimal outcome $A^{*}$, each player chooses exactly one arc from each segment, and players share arcs as little as possible. Straightforward counting based on the above definitions yields that the optimal social cost is

$$
C\left(A^{*}\right)=p \sqrt{x}+3 x^{2}+\left(2 x^{2}\right) 2=p \sqrt{x}+7 x^{2} .
$$

### 6.3. Definition of Strategy Profile $S$ for $\Gamma_{X}$

To describe our worst-case subgame perfect equilibrium strategy, we first define the following actions. Note that these actions are defined relative to a given state in the game. The actions Greedy and Copy are well

Figure 5. A lower bound instance of a network routing game. Players travel from $s$ to $t$.

defined for each state, whereas the actions Punish $(j)$ and Fill exist only for a subset of the states. The actions are as follows:

- Greedy: In each segment, choose the single arc chosen by the least number of players. In cases of ties, the tie-breaking rule as described below is used.
- Punish( $j$ ) (for $j \in[n]$ ): Denote by $R$ a segment where all arcs chosen by player $j$ are chosen by in total less than $x$ players from $[j]$. Denote by $e$ an arc from $R$ that is chosen by the largest number of players among the arcs chosen by $j$, breaking ties in a consistent way. The action $\operatorname{Punish}(j)$ is then defined as choosing $e$ in $R$ and any free arc in each other segment.
- Fill: Choose $\sqrt{x}$ free arcs in each segment.
- Copy: Choose exactly the same arcs as the previous player.

Using these actions, we now define our subgame perfect equilibrium $S=\left(S_{1}, \ldots, S_{n}\right)$ for $\Gamma_{x}$. For each state $\left(A_{<i}, i\right)$, strategy $S_{i}$ prescribes playing an action $S_{i}\left(A_{<i}\right)$, which is determined as in Algorithm 1.

```
Algorithm 1 (Definition of subgame perfect equilibrium \(S\) (for player \(i\) ))
if every player \(j \in[i-1]\) plays according to \(S_{j}\) then
    if \(i\) has at least \(5 x^{2}\) successors then
        if \(i\) is the first player, or if the previous \(\sqrt{x}-1\) players chose Copy then
            Fill
        else
            Copy
        else
            Greedy
        else
        if exactly 1 player \(j \in[i-1]\) does not play according to \(S_{j}\) then
            if \(j\) has chosen less than \(x^{2}\) arcs in each segment then
                    if \(S_{j}\) prescribed \(j\) to choose Fill or Copy then
                    if there exists a segment such that all arcs \(e\) chosen by \(j\) contain less than \(x\) players in total then
                        Punish( \(j\) )
                    else
                        Greedy
        else
            Greedy
```


### 6.4. Tie-Breaking Rule

In Algorithm 1, when the strategy $S_{i}$ prescribes that a player $i$ choose an arc chosen by the largest number of players, any arbitrary fixed tie-breaking rule is used. Second, when the strategy $S_{i}$ prescribes that a player $i$ choose an arch chosen by the smallest number of players, and a set $E^{\prime}$ of multiple arcs have this property, the following tie-breaking rule is used: all predecessors of $i$ are ordered. The set of all players that deviate from $S$ comes first in this ordering. After that comes the set of all other players. Within these two sets, the players are ordered by index from high to low. Now the arcs are ordered as follows: arc $e$ is ordered before $e^{\prime}$ iff the set of players on $e$ is lexographically less than the set of players on $e^{\prime}$ according to the ordering on the players just defined. Finally, ties are broken by choosing the first arc in this order, among the arcs in $E^{\prime}$.
Example 1. As an example to clarify the tie-breaking rule, consider the following situation: Say player 5 has to choose 2 arcs among arc set $\left\{e_{1}, e_{2}, e_{2}, e_{4}\right\}$, which are chosen by the smallest number of players. Players 1 and 3 have deviated from $S$. Player 1 has chosen, among arcs $\left\{e_{1}, e_{2}, e_{3}, e_{4}\right\}$, the two arcs $e_{2}$ and $e_{3}$; player 2 has chosen arcs $e_{3}$ and $e_{4}$; player 3 has chosen $\operatorname{arcs} e_{1}$ and $e_{4}$; and player 4 has chosen $\operatorname{arcs} e_{1}$ and $e_{2}$. Thus, the players are ordered $3,1,4,2$, and the arcs are ordered $e_{4}, e_{1}, e_{3}, e_{2}$, so player 5 chooses arcs $e_{4}$ and $e_{1}$.

Observe first that $S$ is well defined:

- In any state, $S_{i}$ prescribes that $i$ plays either Greedy, Copy, Fill, or Punish( $j$ ) for some $j \in[i-1]$.
- In any state, the actions Greedy and Copy always exist.
- Whenever the action Fill is prescribed, then by line 1, no player in $[i-1]$ has deviated from S. Combined with line 2 , this means that in each segment, there are at most $p \sqrt{x}$ arcs that are chosen by at least one player in $[i-1]$. Therefore, there is guaranteed to be enough free arcs in each segment, so the action Fill exists in any state where $S_{i}$ prescribes $i$ to choose Fill.
- Whenever the action Punish( $j$ ) is prescribed, for some $j \in[i-1]$, then from line 12 and line 2 it follows that $j \in\left[p \sqrt{x}-x^{2}\right]$. Also, from line 10 it follows that in each segment, the total number of arcs chosen by players in $[j-1]$ (who all play Fill and Copy) is at most $p \sqrt{x}-x^{2}+\sqrt{x}<p \sqrt{x}$, and player $j$ chooses at most $x^{2}$ arcs in each segment by line 11. Moreover, it follows from line 10 that players in $\{j+1, \ldots, i-1\}$ have not deviated from $S$ and have therefore chosen only one arc in each segment (as they all have played Punish(j)). From line 13, it follows that the number of players between $j$ and $i$ is at most $x^{2}$, so the total number of arcs chosen in each segment, by players in $\{j+1, \ldots, i-1\}$, is at most $x^{2}$. Last, line 11 certifies that player $j$ occupies at most $x^{2}$ arcs in each segment. So, in a single segment, the total number of arcs used by players in $[i-1]$ is at most $p \sqrt{x}+3 x^{2}$. Thus, when $S_{i}$ prescribes Punish $(j)$, each segment has a free arc. Moreover, by line 13 , there is a segment in which all arcs chosen by player $j$ are chosen by fewer than $x$ players from [ $j$ ]. Therefore, Punish $(j)$ exists in any state where $S_{i}$ prescribes $i$ to choose Punish $(j)$.


### 6.5. Social Cost of $S$

If each player $i$ chooses the action prescribed by $S_{i}$, then the social cost is at least

$$
(p \sqrt{x})(\sqrt{x} \sqrt{x})+3 x^{2}+\left(2 x^{2}\right) 2=p \sqrt{x} x+7 x^{2} .
$$

If we denote by $A$ the outcome induced by $S$, we see that

$$
\lim _{p \rightarrow \infty} C(A) / C\left(A^{*}\right)=\lim _{p \rightarrow \infty}\left(p x \sqrt{x}+7 x^{2}\right) /\left(p \sqrt{x}+7 x^{2}\right)=x .
$$

### 6.6. Checking That $S$ Is a Subgame Perfect Equilibrium

For a state $\left(A_{<i}, i\right)$, an action $A_{i}$ is said to be subgame perfect with respect to a sequential strategy profile $S$ iff choosing $A_{i}$ minimizes $i$ 's cost when players 1 to $i-1$ play $A_{<i}$, and players $i+1$ to $n$ play according to $S$.

The following lemma and corollary shows that $S$ is a subgame perfect equilibrium.
Lemma 1. For each state $\left(A_{<i}, i\right)$ of $\Gamma_{x}$, action $S_{i}\left(A_{<i}\right)$ is subgame perfect with respect to $S$.
Proof. For each of the possible actions, Greedy, Fill, Punish(j) (where $j \in[i-1]$ ), and Copy, that $S_{i}$ may prescribe to player $i$ in state $\left(A_{<i}, i\right)$, we prove that deviating from this prescription will not decrease the cost of player $i$, under the assumption that all succeeding players $i+1, \ldots, n$ play according to $S$ :

- Suppose player $i$ is prescribed by $S_{i}$ to play Fill or Copy. Then no player in $[i-1]$ has deviated from $S$. Therefore, assuming that all succeeding players play according to $S$ as well, the cost of player $i$ when she does not deviate is $x$. If player $i$ does deviate, then the subsequent players will play Punish $(i)$, which ensures that in each
segment, one of the arcs chosen by $i$ gets chosen by at least $x$ players. Her cost will therefore be at least $x$. Thus, deviating is not beneficial for player $i$.
- Suppose player $i$ is prescribed by $S_{i}$ to play Greedy. Then, assuming that players $i+1, \ldots, n$ all play according to $S$ as well, observe that, by definition of $S$, players $i+1, \ldots, n$ play Greedy, even if player $i$ deviates from playing Greedy. We denote by $A$ the outcome that results if $i$ does not deviate from $S_{i}$. We show that if $i$ does deviate, then in each segment, $i^{\prime}$ s costs are at least as high as in $A$. Let $j \in[x]$ and consider segment $R_{j}$. Let $e_{i}$ and $e_{n}$ denote the arcs from $R_{j}$ chosen respectively by player $i$ and player $n$ in $A$. Denote by $R$ the set of arcs in $R_{j}$ chosen by players $i, \ldots, n$ in $A$.
- We denote by $c$ the number of players choosing $\operatorname{arc} e_{n}$ in $A$. Any arc $e \in R$ has cost either $c / x$ or $(c-1) / x$; if it were higher, then the last player who chose $e$ would have chosen $e_{n}$, because she plays greedily. Specifically, the cost of $e_{i}$ is at most $c / x$. Also, any arc $e \in R_{j}$ that is not in $R$ is chosen by at least $c-1$ players of $[i-1]$; if this were false, then in $A$, player $n$ would have chosen $e$ instead of $e_{n}$.
- Now consider outcome $A^{\prime}$, which occurs when player $i$ deviates from $S_{i}$. If player $i$ chooses any arc $e_{i}^{\prime}$ that is not in $R$, then this arc has cost at least $c / x$. We now show that if $e_{i}^{\prime}$ is in $R$, then it has cost at least $c / x$ as well. In that case, if any player $i^{\prime} \in\{i+1, \ldots, n\}$ chooses an arc not in $R$, then all arcs in $R$ will yield cost at least $c / x$. If there were an arc $e^{\prime} \in R$ with $\operatorname{cost}(c-1) / x$, then the tie-breaking rule dictates that $i^{\prime}$ would have chosen $e_{i}^{\prime}$ instead of $e^{\prime}$. However, if all players $i, \ldots, n$ choose an arc in $R$, then player $n$ has cost at least $c / x$. Combining this with the tie-breaking rule, we conclude that $e_{i}^{\prime}$ has a cost of at least $c / x$ as well. Therefore, in all cases, the costs of player $i$ does not decrease by deviating.
- Suppose player $i$ is prescribed by $S_{i}$ to play Punish( $j$ ) for some $j \in[i-1]$. Let us compute first the cost of $i$ if she follows this prescription, assuming that players $i+1, \ldots, n$ all play according to $S$. Then observe that, by definition of $S$, there are a number of other players succeeding $i$ that play Punish $(j)$ as well. Let $k$ be this number of players. So, $\{j+1, \ldots, i+k\}$ is the set of players that play Punish( $j$ ). Let $\ell=|\{j+1, \ldots, i+k\}|$. Players $\{i+k+1, \ldots, n\}$ play Greedy, again by definition of $S$. Players in $[j-1]$ together occupy at most $j-2+\sqrt{x}$ arcs in each segment. Player $j$ occupies at most $x^{2}$ arcs in each segment. Players $j+1, \ldots, i+k$ all choose Punish $(j)$, so they each occupy one arc per segment. The total number of arcs occupied per segment by players in $[i+k]$ is therefore $j-2+x^{2}+\ell+\sqrt{x}$. Therefore, there are at least $F:=(p-1) \sqrt{x}+3 x^{2}-j-\ell+2$ free arcs per segment after the first $i+k$ players have chosen their actions. The set $i+k+1, \ldots, n$ is of size $G:=p \sqrt{x}+5 x^{2}-j-\ell$. We see that

$$
\frac{G}{F}=\frac{p \sqrt{x}+5 x^{2}-j-\ell}{(p-1) \sqrt{x}+3 x^{2}-j-\ell+2} \leq \frac{p \sqrt{x}+6 x^{2}-j-\ell}{(1 / 2) \sqrt{x}+3 x^{2}-(1 / 2) j-(1 / 2) \ell}=2,
$$

so the Greedy players will choose only those free arcs; that is, by the tie-breaking rule, the Greedy players will not choose arcs of player $i$. Therefore, player $i$ 's cost is exactly $2-1 / x$ if she plays $\operatorname{Punish}(j)$ : in $x-1$ segments, $i$ chooses one free arc that will not be chosen by any of her successors, as we have shown. In the remaining segment, $i$ chooses an arc that player $j$ has chosen, which will be chosen by precisely $x$ players.

Suppose now that $i$ deviates from playing Punish( $j$ ). In that case, all succeeding players will play Greedy. We prove that in each segment, $i$ 's costs are at least $2 / x$, so that her total cost is at least 2 . All players in $[j-1]$ together occupy at least $j-1$ arcs per segment. This implies that in state ( $A_{<i}, i$ ) in each segment, there are at least $j-1$ occupied arcs and at most $p \sqrt{x}+4 x^{2}-j$ free arcs. The number of players succeeding $i$ is $p \sqrt{x}+5 x^{2}-$ $i \geq p \sqrt{x}+4 x^{2}-j+x$, where the inequality holds because $i \leq j+x^{2}-x$, which is true because, by the definition of $S$, there are at most $x^{2}-x$ players choosing Punish $(j)$. Therefore, there exist players among the Greedy players who choose in each segment an arc that is occupied by at least one player. The tie-breaking rule for the Greedy action then ensures that the first such a Greedy player chooses in each segment an arc on which $i$ is the sole player, in case such an arc exists. Therefore, when $i$ deviates, her cost in each segment is at least $2 / x$.

Now we can finish the proof of Theorem 5.
Corollary 1. The strategy profile $S$ is a subgame perfect equilibrium of $\Gamma_{x}$.
Proof. Recalling the definition of a subgame perfect equilibrium, we have to prove that for all $i \in[n]$ and for each state $\left(A_{<i}, i\right)$, it holds that the action $S_{i}\left(A_{<i}\right)$ that $i$ plays under $S$ minimizes her cost in the subgame corresponding to state $\left(A_{<i}, i\right)$, when the remaining players play according to $S$. This follows directly from Lemma 1 and from the way we defined subgame perfection of an action with respect to strategy profile $S$.

Although the SPoA is not bounded by any constant, it is not hard to see that it is trivially upper bounded by the number of players, $n$. This holds, as for nonnegative, affine cost functions, $n$ is the maximum multiplicative gap between the congestion of any given resource, for any two arbitrary action profiles. Our construction above shows actually a lower bound SPoA $\geq \Omega(\sqrt{n})$. To see this, we choose $p=x \sqrt{x}$. Then $n=x^{2}+5 x^{2}=6 x^{2}$,
which yields $x=\sqrt{n / 6}$. Now, SPoA $\geq\left(x^{3}+7 x^{2}\right) /\left(x^{2}+7 x^{2}\right) \geq x^{3} /\left(8 x^{2}\right)=x / 8=\sqrt{n} /(8 \sqrt{6})$. Whether instances can be constructed where the sequential price of anarchy is asymptotically worse than $\sqrt{n}$ remains open.

## 7. Discussion

The central result of this paper states that the sequential price of anarchy is generally unbounded for symmetric affine network routing games. One property that stands out in our lower bound constructions is that it admits multiple subgame perfect equilibria. In fact, there even exists a subgame perfect equilibrium that induces an optimal strategy profile, and the existence of a poorly performing subgame perfect equilibrium relies crucially on bad tie breaking. Recent results for special cases, namely, so-called extension-parallel graphs, show that indeed, the sequential price of anarchy is no longer unbounded if we consider generic games, that is, those games admitting a unique subgame perfect equilibrium; see Groenland and Schäfer [16]. We do not know, however, whether the sequential price of anarchy is bounded for all symmetric congestion games that are generic. That question relates to determining the sequential price of stability of symmetric network routing games.

Finally, note that our main result is asymptotic for large $n$. In view of concrete applications, it may be interesting to pin down the exact value for small numbers of $n$; here we did that for $n=2$.

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## Appendix. Proof of Upper Bound for Theorem 2

We here give the missing proof of the $7 / 5$ upper bound for sequentially played affine congestion games with $n=2$ players.
Lemma A.1. The sequential price of anarchy of symmetric affine congestion games is at most $7 / 5$ when there are $n=2$ players only.
Because we consider nonnetwork congestion games for this upper bound proof, we use the term resources instead of arcs. Consider any symmetric affine congestion game with two players, $m$ resources, affine cost functions $\left(c_{1}, \ldots, c_{m}\right)$, and the set $R \subseteq 2^{[m]}$ such that the action set of both players is $R$. We first show that we may assume without loss of generality that all the cost functions are of the form $c_{e}(y)=y$. This is because we can transform any symmetric affine congestion game as follows: First, by scaling all the cost functions simultaneously by an appropriate constant, we may assume that each cost function $c_{e}$ is of the form $c_{e}(y)=a_{e} y+b_{e}$, with $a_{e}$ and $b_{e}$ being natural numbers. (It is straightforward to do so when the original coefficients of the affine cost functions are rational. In case the original coefficients are irrational, additionally one must first turn such an instance into an equivalent one where the coefficients are approximated by rational numbers.) For a resource $e$ where $a_{e} \neq 1$ or $b_{e} \neq 0$, we may replace each action $A^{\prime} \in R$ that contains $e$ by two new strategies, $A^{1}=$ $\left(A^{\prime} \backslash e\right) \cup\left\{e_{j}^{\prime}: j \in\left[a_{e}\right]\right\} \cup\left\{e_{j}^{\prime \prime}: j \in\left[b_{e}\right]\right\}$ and $A^{2}=\left(A^{\prime} \backslash e\right) \cup\left\{e_{j}^{\prime}: j \in\left[a_{e}\right]\right\} \cup\left\{e_{j}^{\prime \prime \prime}: j \in\left[b_{e}\right]\right\}$, where for all $j$, the resources $e_{j}^{\prime}, e_{j}^{\prime \prime}$, and $e_{j}^{\prime \prime \prime}$ are newly introduced into the game, and all these newly introduced resources have cost functions of the form $c_{e}(y)=y$. There is a natural mapping from the old outcomes to the new outcomes: for player 1 , action $A^{\prime}$ is mapped to $A^{1}$, and for player 2 , action $A^{\prime}$ is mapped to $A^{2}$. Note that because the transformed game is still symmetric, action $A^{2}$ can be chosen by player 1, and action $A^{1}$ can be chosen by player 2 . However, the two players will not both choose action $A^{1}$ if $b_{e}>0$ because player 2 could then improve her cost by choosing the action $A^{2}$, and a symmetric conclusion can be made if both players would choose action $A^{2}$. Thus, by this transformation, the sequential price of anarchy can only increase. This shows that we may assume that the cost functions have the form $c_{e}(y)=y$.

Now, let $A^{*}=\left(A_{1}^{*}, A_{2}^{*}\right)$ be an outcome that minimizes the social cost, and let $A=\left(A_{1}, A_{2}\right)$ be the outcome resulting from a subgame perfect equilibrium $S=\left(S_{1}, S_{2}\right)$. We may assume w.l.o.g. that $c_{1}\left(A^{*}\right) \leq c_{2}\left(A^{*}\right)$, because the action sets are symmetric. We first define some parameters that we will use throughout this proof:

- Let $x$ be the minimum cardinality of a set of resources in the action set $R$.
- Let $\alpha \in \mathbb{R}$ be such that $\left|A_{1}^{*} \cap A_{2}^{*}\right|=\alpha x$.
- Let $\beta$ be such that $\left|A_{1}^{*} \cap A_{1}\right|=\beta\left|A_{1}\right|$. Note that this implies that $\left|A_{2}^{*} \cap A_{1}\right| \leq(1-\beta)\left|A_{1}\right|+\alpha x$.
- Let $\gamma$ be such that $C\left(A^{*}\right)=(2+\gamma) x$. Observe that $\gamma$ is positive, as $2 x$ is a lower bound on $C\left(A^{*}\right)$.
- Let $\delta$ be such that $c_{1}\left(A^{*}\right)=(1+\delta) x$. It holds that $\delta \leq \gamma / 2$ because $c_{1}\left(A^{*}\right) \leq c_{2}\left(A^{*}\right)$. The definition of $\delta$ implies that $c_{2}\left(A^{*}\right)=(1+\gamma-\delta) x$.

We derive several upper bounds on $C(A)$ that are expressed in terms of $\alpha, \beta, \gamma$, and $\delta$.
Observe first that each player experiences a cost of at most $2 x$ under $A$, because there exists an action of cardinality $x$ by which each resource is chosen by at most two players.

Proposition A.1. It holds that

$$
c_{1}(A) \leq 2 x, \quad c_{2}(A) \leq 2 x
$$

This gives us a straightforward upper bound of $4 x$ on $C(A)$. Note that in the case $\gamma \geq 6 / 7$, we obtain that

$$
\frac{C(A)}{C\left(A^{*}\right)} \leq \frac{4}{20 / 7}=\frac{7}{5}
$$

so it remains to prove the claim for the case that $\gamma \in[0,6 / 7]$, which we assume from now on.
We prove a second upper bound on $C(A)$ next. By subgame perfection of $S$, it holds that $c_{1}(A) \leq c_{1}\left(A_{1}^{*}, S_{2}\left(A_{1}^{*}\right)\right)$. Also, $c_{2}\left(A_{1}^{*}, S_{2}\left(A_{1}^{*}\right)\right) \leq c_{2}\left(A_{1}^{*}, A_{2}^{*}\right)=(1+\gamma-\delta) x$. The number $x$ is defined as the smallest cardinality of an action in $R$, so $S_{2}\left(A_{1}^{*}\right)$ intersects with $A_{1}^{*}$ in at most $(\gamma-\delta) x$ resources. We combine the latter with the fact that the cardinality of $A_{1}^{*}$ is $(1+\delta-\alpha) x$, and we conclude that $c_{1}\left(A_{1}^{*}, S_{2}\left(A_{1}^{*}\right)\right) \leq(1+\delta-\alpha) x+(\gamma-\delta) x$. Therefore, we have the following proposition:

Proposition A.2. It holds that

$$
c_{1}(A) \leq(1+\gamma-\alpha) x .
$$

Combining this with Proposition A. 1 gives us that

$$
\begin{equation*}
C(A) \leq(1+\gamma-\alpha) x+2 x=(3+\gamma-\alpha) x . \tag{A.1}
\end{equation*}
$$

We prove two additional upper bounds on $C(A)$ next. By subgame perfection of $S$, it holds that $c_{2}(A) \leq c_{2}\left(A_{1}, A_{1}^{*}\right)$, and it holds that $c_{2}(A) \leq c_{2}\left(A_{1}, A_{2}^{*}\right)$. The cost $c_{2}\left(A_{1}, A_{1}^{*}\right)$ can be upper bounded as follows:

$$
\begin{aligned}
\mathcal{c}_{2}\left(A_{1}, A_{1}^{*}\right) \leq\left|A_{1}^{*}\right|+\left|A_{1}^{*} \cap A_{1}\right| & \leq(1+\delta-\alpha) x+\beta\left|A_{1}\right| \\
& \leq(1+\delta-\alpha) x+\beta c_{1}(A) \\
& \leq(1+\delta-\alpha) x+\beta(1+\gamma-\alpha) x
\end{aligned}
$$

where we use Proposition A. 2 for the last inequality. The $\operatorname{cost} c_{2}\left(A_{1}, A_{2}^{*}\right)$ can be upper bounded as follows:

$$
\begin{aligned}
c_{2}\left(A_{1}, A_{2}^{*}\right) \leq\left|A_{2}^{*}\right|+\left|A_{2}^{*} \cap A_{1}\right| & \leq(1+\gamma-\delta-\alpha) x+(1-\beta)\left|A_{1}\right|+\alpha x \\
& \leq(1+\gamma-\delta) x+(1-\beta)\left|A_{1}\right| \\
& \leq(1+\gamma-\delta) x+(1-\beta) c_{1}(A) \\
& \leq(1+\gamma-\delta) x+(1-\beta)(1+\gamma-\alpha) x
\end{aligned}
$$

where we use Proposition A. 2 for the last inequality.
Combining the above two with Proposition A. 2 gives us

$$
\begin{align*}
C(A) & =c_{1}(A)+c_{2}(A) \\
& \leq(1+\gamma-\alpha) x+(1+\delta-\alpha) x+\beta(1+\gamma-\alpha) x \\
& =(1+\delta-\alpha) x+(1+\beta)(1+\gamma-\alpha) x \tag{A.2}
\end{align*}
$$

and

$$
\begin{align*}
C(A) & =c_{1}(A)+c_{2}(A) \\
& \leq(1+\gamma-\alpha) x+(1+\gamma-\delta) x+(1-\beta)(1+\gamma-\alpha) x \\
& =(1+\gamma-\delta) x+(2-\beta)(1+\gamma-\alpha) x . \tag{A.3}
\end{align*}
$$

Combining (A.1), (A.2), and (A.3), we conclude that the sequential price of anarchy of our game is at most

$$
\frac{\min \left\{\begin{array}{c}
(3+\gamma-\alpha) x, \\
(1+\delta-\alpha) x+(1+\beta)(1+\gamma-\alpha) x, \\
(1+\gamma-\delta) x+(2-\beta)(1+\gamma-\alpha) x
\end{array}\right\}}{(2+\gamma) x}=\frac{\min \left\{\begin{array}{c}
(3+\gamma-\alpha), \\
(1+\delta-\alpha)+(1+\beta)(1+\gamma-\alpha), \\
(1+\gamma-\delta)+(2-\beta)(1+\gamma-\alpha)
\end{array}\right\}}{(2+\gamma)} .
$$

We can obtain a concrete upper bound on the sequential price of anarchy of the complete class two-player symmetric affine congestion games when we maximize the latter expression subject to the constraints $\gamma \in[0,6 / 7], \beta \in[0,1], \delta \in[0,1 / 2 \gamma]$. The variable $\alpha$ can be eliminated, as it is clear that the maximum is attained when $\alpha=0$. This results in the optimization problem

$$
\begin{aligned}
\max & \left\{\frac{\min \{(3+\gamma),(1+\delta)+(1+\beta)(1+\gamma),(3-\beta)(1+\gamma)-\delta\}}{(2+\gamma)},\right. \\
& \left.: 0 \leq \gamma \leq \frac{6}{7}, 0 \leq \beta \leq 1,0 \leq \delta \leq \frac{\gamma}{2}\right\} .
\end{aligned}
$$

Numerically solving this program gives that the solution is $7 / 5$, attained when we take $\gamma=\beta=1 / 2, \delta=1 / 4$, although this does not comprise a formal proof. However, it is possible to prove formally that the solution does not exceed $7 / 5$ by showing that the optimal solution to the following optimization problem does not exceed zero:

$$
\begin{align*}
& \max \left\{\min \left\{\begin{array}{c}
(3+\gamma) \\
(1+\delta)+(1+\beta)(1+\gamma) \\
(1+\gamma-\delta)+(2-\beta)(1+\gamma)
\end{array}\right\}-\frac{7}{5}(2+\gamma)\right. \\
& \left.: 0 \leq \gamma \leq \frac{6}{7}, 0 \leq \beta \leq 1,0 \leq \delta \leq \frac{\gamma}{2}\right\} \tag{A.4}
\end{align*}
$$

We introduce an additional variable $z$ that we use to eliminate the min expression in the objective function:

$$
\begin{aligned}
\max & \left\{z-\frac{7}{5}(2+\gamma)\right. \\
& : z \leq 3+\gamma, z \leq(1+\delta)+(1+\beta)(1+\gamma) \\
& z \leq(1+\gamma-\delta)+(2-\beta)(1+\gamma) \\
& \left.0 \leq \gamma \leq \frac{6}{7}, 0 \leq \beta \leq 1,0 \leq \delta \leq \frac{\gamma}{2}\right\}
\end{aligned}
$$

There are still two constraints in this program that are nonlinear, because they contain the terms $\gamma \beta$ and $-\gamma \beta$, respectively. We introduce a new variable $\xi$ that we restrict to lie in $[0,6 / 7]$ and substitute the latter two terms by $\xi$ and $-\xi$, respectively. This results in the following linear program, with an "enlarged" feasible region:

$$
\begin{aligned}
\max & \left\{z-\frac{7}{5}(2+\gamma)\right. \\
& : z \leq 3+\gamma, z \leq 2+\delta+\beta+\gamma+\xi, z \leq 3+3 \gamma-\delta-\beta-\xi, \\
& \left.0 \leq \gamma \leq \frac{6}{7}, 0 \leq \beta \leq 1,0 \leq \delta \leq \frac{\gamma}{2}, 0 \leq \xi \leq 6 / 7\right\}
\end{aligned}
$$

Because the set of feasible points of this linear program is larger, the solution to this linear program is an upper bound to the solution of (A.4). The exact solution to this linear program can be obtained by known algorithms, and turns out to be 0, as we needed to show.

## Endnotes

${ }^{1}$ In a generic game, the players are never faced with a tie, for any strategy profile of the other players. Thus, all decisions are uniquely determined, and in particular, uniqueness of subgame perfect equilibria is guaranteed.
${ }^{2}$ Note that our main results are lower bounds on the price of anarchy and they already apply under linear cost functions.
${ }^{3}$ In de Jong and Uetz [10], a lower bound example is given for general congestion games which can be easily transformed to network routing games.
${ }^{4}$ Just like the price of stability as defined by Anshelevich et al. [2], the sequential price of stability is the ratio of the outcome of the best subgame perfect equilibrium over the optimum.

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