Locally Stable Marriage with Strict Preferences*

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Abstract

We study stable matching problems with locality of information and control. In our model, each agent is a node in a fixed network and strives to be matched to another agent. An agent has a complete preference list over all other agents it can be matched with. Agents can match arbitrarily, and they learn about possible partners dynamically based on their current neighborhood. We consider convergence of dynamics to locally stable matchings – states that are stable with respect to their imposed information structure in the network. In the two-sided case of stable marriage in which existence is guaranteed, we show that the existence of a path to stability becomes NP-hard to decide. This holds even when the network exists only among one partition of agents. In contrast, if one partition has no network and agents remember a previous match every round, a path to stability is guaranteed and random dynamics converge with probability 1. We characterize this positive result in various ways. For instance, it holds for random memory and for cache memory with the most recent partner, but not for cache memory with the best partner. Also, it is crucial which partition of the agents has memory. Finally, we present results for centralized computation of locally stable matchings, i.e., computing maximum locally stable matchings in the two-sided case and deciding existence in the roommates case.

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1 Introduction

Matching problems form the basis for a variety of assignment and allocation tasks encountered in computer science, operations research, and economics. A prominent and popular approach in all these areas is *stable matching*, as it captures aspects like distributed control and rationality of participants that arise in many assignment problems today. A variety of allocation problems in markets can be analyzed within the context of two-sided stable matching, e.g., the assignment of jobs to workers [5, 28], organs to patients [36], or general buyers to sellers. In addition, stable marriage problems have been successfully used to study distributed resource allocation problems in networks [3,17,33].

In this paper, we consider a game-theoretic model for decentralized matching with limited information. A set of rational agents are embedded in a (social) network and strive to find a partner for a joint relationship or activity, e.g., to do sports, write a research paper, exchange data etc. Such problems are of central interest in economics and sociology, and they act as fundamental coordination tasks in distributed computer networks. Our model extends the stable marriage problem, in which we have sets A and B of men and women. Each man (woman) can match to at most one woman (man) and has a complete preference list over all women (men). Each agent would rather be matched than unmatched. Given a matching M, a blocking pair is a man-woman pair such that both improve by matching to each other and leaving their current partner (if any). A matching without any blocking pair is a stable matching.

A central assumption in stable marriage is that every agent knows all agents it can match to. In reality, however, agents often have limited information about their matching possibilities. For instance, in a large society we would not expect a man to match up with any other woman immediately. Instead, there exist restrictions in terms of knowledge and information that allow some pairs to match up directly, while others would have to get to know each other first before being able to start a relationship. Similar information restrictions exist in many large matching markets – in job markets, one of the most successful strategies to find a new job is to rely on personal contacts that allow the discovery of job opportunities before they become public knowledge. In housing markets, a popular strategy for finding a new roommate is to rely on personal contacts in the social network to find possible tenants. In dating markets, agents usually get to know only a small subset of agents based on which further contacts can evolve. The same happens in many matching scenarios, e.g., for finding partners to travel or to do sports.

We incorporate this aspect of local information by assuming that agents are embedded in a fixed network of $links\ L$. Links represent an enduring knowledge relation that is not primarily under the control of the agents. Depending on the interpretation, links could represent, e.g., family, neighbor, co-worker or teammate relations. Each agent strives to build one $matching\ edge$ to a partner. The set of links and edges defines a dynamic information structure based on $triadic\ closure$, a standard idea in social network theory: If two agents have a common friend, they are likely to meet and learn about each other. Translated into our model this implies that each agent can match to partners in its 2-hop neighborhood of the network of matching edges and links. To clarify our discussion, we present a formal definition here. For more preliminaries and details on the model see Section 2.1.

Definition 1 (Locally Stable Matching). We denote by V the set of agents, and $E \subset V \times V$ the set of possible matching edges. A local blocking pair of matching M is a blocking pair $\{u, v\} \in E$ of agents $u, v \in V$ which are at hop distance at most 2 in the network $G = (V, M \cup L)$. Consequently, a locally stable matching is a stable matching without local blocking pairs.

Local blocking pairs are a subset of blocking pairs. This implies that every stable matching is a locally stable matching, because it allows no (local or global) blocking pairs. Thus, one might

be tempted to think that locally stable matchings are easier to find and/or reach using distributed dynamics than ordinary stable matchings. In contrast, we show in this paper that locally stable matchings have a rich structure and can behave quite differently than ordinary stable matchings. Our study of locally stable matching with general strict preferences significantly extends recent work on the special case of correlated preferences [19, 20, 22], in which preferences are determined by benefits for each matching edge.

For most of the paper, we concentrate on the important two-sided scenario of stable marriage, in which a (locally) stable matching is always guaranteed to exist. Our primary interest is to characterize convergence properties of iterative round-based dynamics with distributed control, in which in each round a local blocking pair is resolved. We focus on the PATH-TO-STABILITY problem:

Definition 2 (PATH-TO-STABILITY). Given a local matching game and an initial matching M_0 , the problem is to decide if there is a path to stability, i.e., a sequence of local blocking pair resolutions leading from M_0 to an arbitrary locally stable matching. As a variation, we sometimes consider existence of a path to a given locally stable matching instead of an arbitrary one.

The properties of such dynamics in large matching markets are a broad domain of interest in the literature [16,35] (see also our discussion of related work below). We show that, in contrast to stable matching without locality restrictions, PATH-TO-STABILITY can be NP-hard, even in surprisingly special cases. There are two conditions we identify to overcome this lower bound – correlated structure in the preference lists, and memory.

Memory is a natural and important characteristic of human behavior. Moreover, even when agents are, e.g., organizations, or software/hardware components, it is reasonable to assume a certain form of memory. We take a quite conservative approach to this issue and assume that in each step of a sequence, each agent can remember only a single other agent that it was matched to before. This single agent can be chosen in various ways, and we consider three natural strategies – the best partner (quality memory), the most recent partner (recency memory), a random partner (random memory). Our results show that the right kind of memory can indeed drastically change the properties of PATH-To-STABILITY. In fact, perhaps the most effective strategy of the three is random memory, since it allows to guarantee convergence in the limit with probability 1. However, recency memory also can be effective under some restrictions on the structure of the instance. Quality memory, however, does not seem to significantly change the complexity of PATH-To-STABILITY.

1.1 Results and Contribution

In Section 3 we first derive an example instance, where a locally stable matching can never be reached when starting from the empty matching. This is in strong contrast to the case of correlated preferences (sometimes referred to as weighted matching, or globally ranked pairs), in which it is easy to show convergence of every sequence of local blocking pair resolutions with a potential function [33]. In fact, we use this gadget to show that it is NP-hard to decide PATH-TO-STABILITY, even if there are no links within one partition of agents. If we need to decide PATH-TO-STABILITY from a given initial matching to a given locally stable matching, we show that this is NP-hard even for correlated preferences, and even if the links are only within one partition of agents.

Moreover, we prove that there exist games and initial matchings such that *every* sequence of local blocking pair resolutions terminating in a locally stable matching is exponentially long. Hence, in general PATH-TO-STABILITY might even be outside NP. In contrast, for correlated preferences, we show that every reachable state can also be reached using a sequence of polynomial length. Thus, for correlated preferences, PATH-TO-STABILITY is in NP and, thus, NP-complete if we ask

for a given locally stable matching¹.

Given our NP-hardness results, in Section 4 we concentrate on a more restricted class of games in which one partition has no internal links, i.e., links exist only between partitions and among the other partition. This is a natural assumption when considering objects that do not generate knowledge about each other, e.g., when matching resources to networked nodes or users, where initially resources are only known to a subset of users. Here we characterize the impact of memory on distributed dynamics. For recency memory, each agent remembers in every round the most recent partner that is different from the current one. With recency memory, we show that PATH-TO-STABILITY is always true, and for every initial matching there exists a sequence of polynomially many local or remembered blocking pairs leading to a locally stable matching. In fact, we only need the partition without internal links to have recency memory. If, in contrast, only the other partition has recency memory, PATH-TO-STABILITY becomes again NP-hard. The same hardness holds for quality memory if all agents from both partitions remember their best partner. Our results formally support the intuition that recency memory is more powerful than quality memory, as the latter can be easily misled in the course of a dynamic process. This provides a novel distinction between recency and quality memory that was not known in previous work [19].

Our positive results for recency memory in Section 4 imply that if we pick admissible blocking pairs uniformly at random in each step, we achieve convergence with probability 1. This can also be guaranteed for *random memory* if in each round each agent remembers one of his previous matches chosen uniformly at random. In fact, for random memory this result holds even in general when links exist among or between both partitions. However, using known results on stable marriage with full information [2], convergence time can be exponential with high probability, independently of any memory.

In Sections 5.1 and 5.2 we treat more centralized aspects of locally stable matching to highlight their different nature compared to ordinary stable matchings. A fundamental observation that motivates our results in Section 5.1 is that – in contrast to ordinary stable matchings – two locally stable matchings can have different sizes, and we consider the natural problem of finding a locally stable matching of maximum cardinality. This problem is known to be APX-hard [14]. While a simple 2-approximation algorithm exists, we can show a non-approximability result of $1.5 - \varepsilon$ under the unique games conjecture. Finally, in Section 5.2 we consider the roommates problem, in which agents can match arbitrarily to other agents. In this case, we show that – in contrast to ordinary stable matchings – deciding existence of locally stable matchings is NP-complete.

Note that we prove all our hardness results for the case when agents have complete lists. In contrast, all our positive results are shown for the case of incomplete lists, where for each agent the set of possible matching partners is restricted to some arbitrary subset of agents (or, in case of stable marriage, subset of the other partition).

1.2 Related Work

Locally stable matchings were introduced by Arcaute and Vassilvitskii [5] in a two-sided job-market model, in which links exist only among one partition. The paper uses strong uniformity assumptions on the preference lists and addresses the lattice structure for stable matchings and a local Gale-Shapley algorithm.

More recently, we studied locally stable matching and extensions with correlated preferences [19]. In the roommates problem, where arbitrary pairs of agents can be matched, a potential function argument shows that PATH-TO-STABILITY is always true and convergence guaranteed [1,34]. Moreover, in [19] we proved that for every initial matching there is a polynomial path to stability, i.e., a

¹Note that it is always true if we ask for an arbitrary one [19].

sequence of local blocking pairs that leads to a locally stable matching. The expected convergence time of random dynamics, however, can be exponential. If we restrict to resolution of pairs with maximum benefit, then for random memory the expected convergence time becomes polynomial, but for recency or quality memory convergence time remains exponential, even if the memory is of polynomial size.

Subsequent to publication of the extended abstract of this paper in [21], we studied more general coalition formation games with correlated preferences in which a polynomial path to stability can always be guaranteed. In [22] we extended some of the results shown here (the NP-hardness result in Theorem 2 and the existence result in Theorem 4) to other variants of matching games, such as considerate or friendship matching [4] that capture externalities among agents. In addition, we provided a tight characterization for a class of games with more general dynamic restrictions on the set of available blocking coalitions. This class contains locally stable matching as well as a variety of other variants of matching games. In [20] we showed a tight characterization for graph-based coalition formation games with limited visibility. This class extends the ideas underlying locally stable matching to coalitions of larger size.

For ordinary two-sided stable matching there have been a wide variety of works on various aspects, e.g., many-to-many matchings, ties, incomplete lists, etc. For an introduction to the topic we refer the reader to several books in the area [18,31,37]. Theoretical work on convergence issues in ordinary stable marriage has focused on better-response dynamics, in which agents sequentially deviate to blocking pairs. It is known that for stable marriage these dynamics can cycle [30]. On the other hand, PATH-TO-STABILITY is always true, and for every initial matching there exists a polynomial path to stability to an arbitrary stable matching [38]. Our recent work [22] shows that deciding PATH-TO-STABILITY to a given stable matching becomes NP-hard. If blocking pairs are chosen uniformly at random at each step, convergence time to an arbitrary stable matching can be exponential [2]. More recently, several works studied convergence time of random dynamics using combinatorial properties of preferences [23], or the probabilities of reaching certain stable matchings via random dynamics [9]. A prominent open problem in this domain is deciding PATH-TO-STABILITY for a sequence of blocking switches, where if a blocking pair $\{a,b\}$ deviates, then their former partners also pair up immediately. A path to stability does not necessarily exist [39,40], but up to our knowledge the complexity of deciding PATH-TO-STABILITY is still open.

For the more general roommates problem, in which every pair of agents can be matched, stable matchings can be absent. There are algorithms to decide existence and compute stable matchings in polynomial time if they exist [26]. Some of these algorithms rely on a tight characterization of stable matchings in terms of linear programming [41]. In terms of convergence, if a stable matching exists, there also exist polynomial paths to stability to a stable matching from every initial matching [15]. Similar results hold for more general concepts like P-stable matchings that always exist [24]. Ergodic sets of the underlying Markov chain have been studied [25] and related to random dynamics [29]. Alternatively, several works have studied the computation of (variants of) stable matchings using iterative entry dynamics [8, 10, 11, 13].

Recently, computing locally stable matchings with maximum cardinality has attracted some interest. While local algorithms perform arbitrarily badly [12], the problem was shown to be APX-hard in [14] and non-approximable within 21/19 unless $P \neq NP$. We show a stronger lower bound of $1.5 - \varepsilon$ under unique games conjecture. The same question has been studied for a related variant termed socially stable matching. Askalidis et al [6] show that the problem in this variant is also non-approximable within $1.5 - \varepsilon$ under the unique games conjecture by adapting the proof technique of this paper. They also provide a 1.5-approximation algorithm for socially stable matchings.

2 Preliminaries

2.1 Locally Stable Matching

A network matching game (or network game) consists of a (social) network N=(V,L), where V is a set of vertices representing agents and $L\subseteq \{\{u,v\}\mid u,v\in V,u\neq v\}$ is a set of fixed links. The set $E\subseteq \{\{u,v\}\mid u,v\in V,u\neq v\}$ defines the potential matching edges. Note that, in general, E is not necessarily a sub- or superset of E. A state E is a matching, where for each E is have $|\{e\mid e\in M,v\in e\}|\leq 1$. Based on a state, we define the graph to be E to be

If an edge $e = \{u, v\} \in M$, the incident agents obtain a utility of $b_u(e), b_v(e) > 0$ for u and v, respectively. Alternatively, we often represent the preference of each agent v by a list \succ_v over all its possible matching partners, where $u \succ_v w$ if $b_v(\{v, u\}) > b_v(\{v, w\})$ and $u =_v w$ if $b_v(\{v, u\}) =_v b_v(\{v, w\})$, for any $\{v, u\}, \{v, w\} \in E$. Note here that our model includes instances with incomplete lists and ties, and all our upper bounds hold for this general case. In contrast, our lower bounds apply even for the case of strict preference lists. As a special case, we also study games with correlated preferences (also termed correlated network games), in which it holds that $b_u(e) = b_v(e) = b(e) > 0$ for every $e \in E$.

If we can divide V into two disjoint sets U and W such that $E \subseteq \{\{u, w\} \mid u \in U, w \in W\}$, we call the game *bipartite*. We will focus on this case in Section 3 and Section 4. Note that this does not imply that N has to be bipartite. If further the agents of U are isolated in N, we term the game a *job-market game* for consistency with [5,19].

To describe stability in network matching games, we extend the classic notion of blocking pair. Consider a matching M and the graph G_M . A possible matching edge $e = \{u, v\} \in E$ is termed preferred by u in M if u is either (1) unmatched in M or (2) matched by edge $\{u, w\} \in M$ such that $v \succ_u w$. The edge $e \in E$ is termed a blocking pair in M if it is both preferred by u and v. Further, we term the pair $\{u, v\}$ accessible in M if u and v are connected via a path of length at most 2 in G_M . A possible matching edge $e = \{u, v\} \in E$ is termed a local blocking pair in M if it is both a blocking pair and accessible in M. Thus, for a local blocking pair e both agents can strictly increase their utility by deviating jointly to e (and dismissing any existing incident edges). A state M without a local blocking pair is termed a locally stable matching.

Most of our analysis concerns iterative round-based dynamics. In this process, we pick in each step one local blocking pair $e = \{u, v\}$, remove all edges incident to u and v in M, and then add e to M. We call one such step a local improvement step. By random dynamics we refer to the process when in each step the local blocking pair is chosen uniformly at random from the ones available. Consider a local blocking pair $\{u, v\} \in E$ that is resolved in such a step. Before the step, u and v are either connected directly by a link $\{u, v\} \in L$, by a path of two links $\{u, w\}, \{w, v\} \in L$, or by a path of a single matching edge and a single link in L. In the latter case, let w.l.o.g. $\{u, w\}$ be the matching edge and $\{v, w\}$ the link. Since $\{u, w\}$ is the only edge of M incident to u, the local improvement step will remove $\{u, w\}$ to create $\{u, v\}$. This observation will be helpful in our constructions below, and for simplicity we will refer to such a step as "an edge moving from $\{u, w\}$ to $\{u, v\}$ " or "u's edge moving from w to v".

In subsequent sections, we will consider dynamics in which each agent has a memory that allows to "remember" one matching partner from a previous step. Memory can be thought of as a cache of size 1 to store a single previous partner. In this case, we extend the notion of accessible agents as follows. A pair $\{u, v\}$ of agents becomes accessible in M if (1) there is a u-v-path of distance at most 2 in G_M , or (2) if u is present in the memory of v, or v in the memory of u. All other definitions follow accordingly. Hence, in this case a local blocking pair can be based solely on access through memory. We consider three strategies for memory update. For v-random v-ra

Agent	Preference List
1	$C \succ B \succ A$
2	$A \succ C \succ B$
3	$B \succ A \succ C$
A	$3 \succ 1 \succ 2$
B	$1 \succ 2 \succ 3$
C	$2 \succ 3 \succ 1$

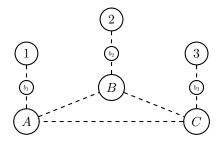


Figure 1: Preferences and links in the circling gadget.

that in every step the memory contains a previous matching partner chosen uniformly at random. For *recency memory*, each agent keeps in memory the last matching partner that is different from the current partner. For *quality memory*, each agent remembers the previous matching partner that gave him the highest utility.

3 Reachability in Bipartite Network Games

In this section we focus on lower bounds for the PATH-TO-STABILITY problem in bipartite network games. Throughout, we focus on the empty matching as a natural candidate for the initial matching and show that PATH-TO-STABILITY is NP-hard to decide. This is in contrast to correlated network games, where PATH-TO-STABILITY is always true, and for every initial matching there is a polynomial path to a locally stable matching [19]. However, we show that given a distinct matching to reach, deciding PATH-TO-STABILITY becomes NP-hard, even for correlated job-market games.

Additionally, we give a network game and an initial matching such that we need an exponential number of steps before reaching any locally stable matching. This is again in contrast to the correlated case, in which we show that every reachable locally stable matching can be reached by a polynomial path.

3.1 Complexity

Our constructions rely on a simple instance, in which it is impossible to reach a locally stable matching from the empty initial matching. We term this structure a *circling gadget*.

Example (Circling Gadget). The agent set V consists of the classes $U = \{1, 2, 3\}$ and $W = \{A, B, C, b_1, b_2, b_3\}$. The links are $L = \{\{A, b_1\}, \{B, b_2\}, \{C, b_3\}, \{1, b_1\}, \{2, b_2\}, \{3, b_3\}, \{A, B\}, \{B, C\}, \{C, A\}$ (see Fig. 1 below). For simplicity, we restrict the possible matching edges to $E = \{\{u, v\} \mid u = 1, 2, 3, v = A, B, C\}$. It will become obvious that by ranking b_1, b_2, b_3 at the bottom, we can similarly allow $E = U \times W$ without changing the arguments.

This gadget has two locally stable matchings, namely

$$\{\{1,B\},\{2,C\},\{3,A\}\}$$
 and $\{\{1,C\},\{2,A\},\{3,B\}\}$.

However, we will show that from every state in which one agent of W is unmatched, every possible sequence of local improvement steps leads to some state where some agent of W is unmatched again. The symmetry of the gadget then allows to repeat the argument, and thus no locally stable matching can be reached. For our analysis, let us w.l.o.g. assume that 1 is unmatched.

First, suppose A is not matched to 3. Then $\{A,1\}$ is a local blocking pair, since the agents are at hop distance 2 from each other, 1 is unmatched and A prefers 1 over 2. When creating this

edge, the distance from 1 to B reduces from 3 to 2. Agent 1 is the top choice of B. Hence, as long as 1 remains matched to A, $\{B,1\}$ remains a local blocking pair. When A gets matched to some other agent, then 1 becomes unmatched. Otherwise, if we create $\{B,1\}$, A is left unmatched, which implies that also one of the agents 2 or 3 must be unmatched.

Now suppose A is matched to 3. Following the arguments above, as long as B is matched to 2, $\{C,2\}$ is a local blocking pair. Hence, suppose that B is not matched to 2. Since 1 is unmatched, $\{B,3\}$ is a local blocking pair. If we create this edge, A becomes unmatched, and we reach the case analyzed above. Hence, every sequence of moves yields a state in which another agent of $\{1,2,3\}$ is unmatched.

In turn, it is simple to verify that when agent A is matched to some more preferred partner outside of the gadget, the remaining agents can always stabilize easily through local improvement steps. For example, starting from the empty matching, consider the following sequence: $\emptyset \to \{(2,B)\} \to \{(2,C)\}$. The latter is a locally stable matching when A is matched to a preferred partner outside the gadget.

Finally, observe that all the previous arguments can be made if $E = U \times W$ and the agents b_i are ranked arbitrarily at the bottom of the lists of agents in U. Consider agent b_1 , the argument is similar for all other agents b_i . b_1 is only accessible with agent $i' \in U$, $i' \neq 1$, when i' is matched to A. At this point, however, i' has no incentive to deviate to b_1 . Hence, each b_i can only be involved in the local blocking pair $\{i, b_i\}$. Since these edges also exist as links, their creation does not change any of the arguments made above.

In our hardness proofs we use this gadget if we need to force a certain agent to be matched in a locally stable matching. For this we identify the forced agent with A of the gadget and declare all allowed outside connections preferable to the gadget agents. Then, if the agent is matched to a partner outside the gadget, the gadget can stabilize while otherwise it does not.

Our first theorem proves NP-hardness of deciding PATH-TO-STABILITY to a given locally stable matching. In our related work, we provided a simpler proof template for this statement in a variety of matching games with additional visibility or externality constraints [22]. The template, however, does not work for the special case of a job-market game (i.e., links exist only within one partition) with correlated preferences (i.e., preferences are correlated by a single value b(e) for each matching edge). Since the empty matching is always locally stable in job market games, we use a different initial state to show the result.

In addition, the proof will introduce the main construction and intuition that will be used to show a number of hardness proofs for different scenarios with and without memory throughout the paper.

Theorem 1. It is NP-hard to decide PATH-TO-STABILITY to a given locally stable matching in a correlated job-market game.

Proof. We reduce from 3SAT. Let the given 3SAT-formula contain k variables x_1, \ldots, x_k and l clauses C_1, \ldots, C_l , where clause C_j holds the literals l_{1j}, l_{2j} and l_{3j} . For each variable x_i , our instance contains an agent u_{x_i} in U, and four agents $a_{x_i}, x_i, \overline{x}_i$ and v_{x_i} in W. For every clause C_j , our instance contains an agent u_{C_j} in U and two agents a_{C_j} and v_{C_j} in W. Further, there is an agent a in W. We allow all possible matching edges $E = U \times W$, but due to the structure of the instance and our choice of initial state, some of these edges can never be created during the dynamics.

First, let us describe the link structure. Since we have a job market game, links are all among agents in W. The network N forms a path with a branching in the middle, see Fig. 2. The first part of the path contains links $\{a_{C_j}, a_{C_{j+1}}\}$ for $j = 1, \ldots, l-1$, a link $\{a_{C_l}, a_{x_1}\}$, links $\{a_{x_i}, a_{x_{i+1}}\}$ for $i = 1, \ldots, k-1$, and a link $\{a_{x_k}, a\}$. The branching is created by links $\{a, x_i\}$, $\{a, \bar{x}_i\}$, $\{x_i, v_{x_1}\}$

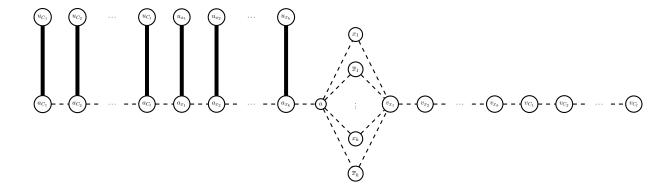


Figure 2: Network and initial matching in the instance for Theorem 1

and $\{\bar{x}_i, v_{x_1}\}$ for i = 1, ..., k. Finally, we obtain the second part of the path by links $\{v_{x_j}, v_{x_{j+1}}\}$ for j = 1, ..., l-1, a link $\{v_{x_k}, v_{C_1}\}$, and links $\{v_{C_j}, v_{C_{j+1}}\}$ for j = 1, ..., l-1.

The initial matching M_0 consists of $M_0 = \{\{u_s, a_s\} \mid s \in \{x_1, ..., x_k\} \cup \{C_1, ..., C_l\}\}$. The given locally stable matching to be reached is $M = \{\{u_s, v_s\} \mid s \in \{x_1, ..., x_k\} \cup \{C_1, ..., C_l\}\}$. In Table 1 we provide a description of the preferences w.r.t. the relevant matching edges in the dynamics. All other matching edges not mentioned in the table can be considered to have edge weight of, say, $\varepsilon \ll 1$. Thereby, these matches are ranked arbitrarily at the bottom of the preference list of each involved agent.

Let us first provide a broad intuition of the dynamics in the gadget. Consider the display of the gadget in Fig. 2. Agents in U will generally prefer to be matched "further to the right" along the path in N (i.e., most preferred v_{C_l} , least preferred a_{C_1}). We assume all agents x_i and \bar{x}_i in the branching are equally preferred for each agent $u \in U$, but in order to avoid ties in the preference lists we can also break them arbitrarily. Agents in W generally prefer to have a partner that is "further to the right" in the order the agents of U are displayed (i.e., most preferred u_{x_k} , least preferred u_{C_1}). As such there is, e.g., exactly one local blocking pair for the initial matching, namely $\{u_{x_k}, a\}$.

Since the instance is a job-market game, if an agent in U becomes unmatched, it can never be involved in any local blocking pair and thus stays unmatched from that point on. However, since in M all agents of U are matched, they must stay matched throughout the sequence. Therefore, the sequence will proceed from M_0 by "shifting the edges along the path" – for every $s \in \{x_1, \ldots, x_k, C_1, \ldots, C_l\}$, agent u_s will become matched, one by one, to the agents of W along the left side of the path leading from a_s to a, one agent in the branching, and then the agents on right side leading from the branching to v_s . This is the only way u_s will be able to discover and match to v_s via resolution of local blocking pairs. Hence, the matching edge incident to u_s will "move" along the path, through the branching and then to v_s .

We now describe some adjustments to the general outline above that serve to create a correct reduction. If agent u_{x_i} is matched to a, we restrict his preference so that the only local blocking pairs involving u_{x_i} are $\{u_{x_i}, x_i\}$ or $\{u_{x_i}, \bar{x}_i\}$. Hence, at this point u_{x_i} is only interested in deviating to designated agents x_i, \bar{x}_i . Matching edges $\{u_{x_i}, s\}$ for $s \in \{x_j, \bar{x}_j\}$ with $i \neq j$ become ranked arbitrarily at the bottom of the preference list of the corresponding agents. A similar adjustment is done for the clause agents. If agent u_{C_j} is matched to a, we restrict his preference so that the only local blocking pairs are $\{u_{C_j}, l_{j1}\}$, $\{u_{C_j}, l_{2j}\}$ and $\{u_{C_j}, l_{3j}\}$. Hence, at this point u_{C_j} is only interested in deviating to agents that correspond to literals that evaluate the formula to true.

Furthermore, we "stop" the movement of the matching edge incident to u_{x_i} once it has reached to v_{x_i} – we assume edges $\{u_{x_i}, v_{x_j}\}$ for j > i and $\{u_{x_i}, v_{C_j}\}$ for $j = 1, \ldots, l$ are ranked at the

$u \in U$	$w \in W$	edge weight $b(\{u, w\})$	
u_{C_j}	$a_{C_{j'}}$	2(k+l+1)j+j'	$j, j' \in [l]$
u_{C_j}	a_{x_i}	2(k+l+1)j+l+i	$i \in [k], j \in [l]$
u_{x_i}	a_{C_j}	2(k+l+1)(l+i)+j	$i \in [k], j \in [l]$
u_{x_i}	$a_{x_{i'}}$	2(k+l+1)(l+i) + l + i'	$i, i' \in [k]$
u_{C_j}	a	2(k+l+1)j + k + l + 1	$j \in [l]$
u_{x_i}	a	2(k+l+1)(l+i) + k + l + 1	$i \in [k]$
u_{C_j}	$l_{1j}/l_{2j}/l_{3j}$	2(k+l+1)j + k + l + 2	$j \in [l]$
u_{x_i}	x_i/\bar{x}_i	2(k+l+1)(l+i) + k + l + 2	$i \in [k]$
u_{C_j}	$v_{x_{i'}}$	2(k+l+1)j + (k+l+2) + i'	$i \in [k], j \in [l]$
u_{x_i}	$v_{x_{i'}}$	2(k+l+1)(l+i) + (k+l+2) + i'	$i, i' \in [k], i' \le i$
u_{C_j}	$v_{C_{j'}}$	2(k+l+1)j + (k+l+2) + k + j'	$j, j' \in [l], j' \le j$

Table 1: Preferences of agents in the instance for Theorem 1. Since preference lists are correlated, we here specify the edge weight for a significant subset of matching edges. Agents rank their possible partners in non-increasing order of edge weight. For each edge we did not specify here, we assume there is a very small weight. Edges of small weight are then ranked in arbitrary order at the bottom the preference lists of the incident agents. They do not change the dynamics as described in the proof.

bottom of the lists. A similar adjustment is made to edges $\{u_{C_j}, v_{C_{j'}}\}$ for j' > j. Observe that these adjustments imply that M is a locally stable matching. More precisely, M becomes the unique (globally) stable matching.

Now it is rather simple to see the idea of the reduction. The key property is that once any edge $\{u_{x_i}, v_{x_i}\}$ exists, every agent u_{C_j} must be matched to $v_{x_{i'}}$ with i' > i or $v_{C_{j'}}$. Assume otherwise, then the edge incident to some u_{C_j} must still pass through v_{x_i} to reach v_{C_j} . Since v_{x_i} is the most preferred partner of u_{x_i} , the latter would have to become (and remain) unmatched, at which point M could not be obtained. By similar arguments, M can only be reached if and only if every agent v_{x_i} becomes matched to $u_{C_l}, \ldots, u_{C_1}, u_{x_k}, \ldots, u_{x_i}$ in that order.

Finally, for the formal correctness proof, suppose the 3SAT formula is satisfiable. Then for each variable x_1, \ldots, x_k in that order, we match each x_i iteratively along the path to a and then match it to x_i (\bar{x}_i) if x_i is set false (true) in the satisfying assignment. This leaves unmatched the agents of W in the branching that correspond to the satisfying assignment. Now, for each clause C_l, \ldots, C_1 in this order, we consider agent u_{C_j} , iteratively match it along the path to a, then to one of the literal agents l_{1j}, l_{2j}, l_{3j} (one is unmatched since the assignment is satisfying), and further to v_{C_j} . For each variable x_k, \ldots, x_1 in that order, we then match u_{x_i} from its current partner to v_{x_1} and further to v_{x_i} . Thereby we obtain a sequence from M_0 to M.

In turn, suppose there is a sequence from M_0 to M. Due to the key property, we must match every u_{x_i} to one of its designated agents in the branching before matching any of them to v_{x_1} . This induces an assignment of the variables. Then, since u_{C_j} considers deviating from a only to agents that correspond to its literals, each u_{C_j} must be matched to one of its literal agents in order to become accessible to v_{C_j} . Therefore, to construct the desired sequence, the 3SAT formula must be satisfiable.

The main property of the previous proof is that, intuitively, we need to carefully match the variable agents of U in a branching in order to allow the clause agents of U reach their desired partner. We now drop the requirement of a job-market game and allow links among both partitions

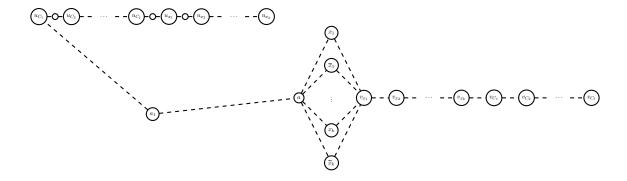


Figure 3: Network in the instance for Theorem 2

of the agents. Thereby, we can even establish the same result starting from the empty matching. This is again a stronger result than the one derived by the template in [22], since the template does not apply to an empty initial matching.

Theorem 2. It is NP-hard to decide PATH-TO-STABILITY from the initial matching $M = \emptyset$ to a given locally stable matching in a correlated bipartite network game.

Proof. The argument will heavily rely on the ideas put forward in the previous proof. The main adjustment here is that we replace the left side of the path (agents a_{x_i} and a_{C_j} , for $i \in [k], j \in [l]$) by a different construction that allows to create matching edges for unmatched agents in U. In contrast to the previous construction, we add agents b_h , $h = 1, \ldots, l + k - 1$, to U, and replace all agents a_{x_i} and a_{C_j} by a single agent a_1 in W. For the link set L, consider the display of the instance in Fig. 3. The b-agents are not labeled in the picture as they only act as a buffer to ensure that local blocking pairs involving b_h exist only when the neighboring u-agent is matched to a. We again construct a path with a branching in a similar fashion as above. For the left side of the path, we add links $\{a, a_1\}, \{a_1, u_{C_1}\}, \{u_{C_j}, b_j\}$ for $j = 1, \ldots, l$, $\{b_j, u_{C_{j+1}}\}$ for $j = 1, \ldots, l - 1$, $\{b_l, u_{x_1}\}, \{u_{x_i}, b_{l+i}\}$ for $i = 1, \ldots, k - 1$ and $\{b_{l+i}, u_{x_{i+1}}\}$ for $i = 1, \ldots, k - 1$. We start from the empty matching, and the the goal is again to reach $M = \{\{u_s, v_s\} \mid s \in \{x_1, ..., x_k\} \cup \{C_1, ..., C_l\}\}$. For a formal description of the correlated preference lists, see Table 2.

Suppose the 3SAT formula is satisfiable. We describe the sequence to reach M, which shows the general idea of the construction. First, we match a iteratively to $u_{C_1}, b_1, u_{C_2}, b_2, \ldots, b_{k+l-1}, u_{x_1}$. Then, we match ux_1 to x_1 (\bar{x}_1) if x_1 is false (true) in the satisfying assignment. Afterwards, we match a again iteratively to $u_{C_l}, b_1, u_{C_{l-1}}, b_2, \ldots, b_{k+l-2}, u_{x_2}$, and then u_{x_2} to x_2 (\bar{x}_2) if x_2 is false (true) in the satisfying assignment. In similar fashion, we proceed with all variable agents. Once u_{x_k} is matched to either x_k or \bar{x}_k , we proceed with the clause agents. We match a iteratively to $u_{C_1}, b_1, u_{C_2}, b_2, \ldots, b_{l-1}, u_{C_l}$, match u_{C_l} to an agent corresponding to a literal that evaluates C_l to true, and iteratively further down the path to v_{C_l} . In similar fashion, we proceed with $u_{C_{l-1}}, \ldots, u_{C_1}$. Finally, we match u_{x_i} to v_{x_1} and down the path to v_{x_i} , for each $i = k, \ldots, 1$ in that order.

Note that agent a and the path among agents u_s and b_h allow to create matching edges for all agents u_s and shift them iteratively to agents in the branching, for u_s with $s = x_1, \ldots, x_k, C_l, \ldots, C_1$ in that order. In particular, the construction requires that edges incident to x_1, \ldots, x_k have to be created before any of u_{C_1}, \ldots, u_{C_l} is matched to an agent in the branching. Otherwise, a key property similar to the one in the previous proof implies that the desired matching cannot be reached. This is the main intuition in the following proof of the reverse direction.

Suppose there is a sequence to reach M from the empty matching. With the exception of u_{C_l} , none of the agents from u_{C_i} or u_{x_i} are at hop-distance 2 from any agent of W in the empty matching.

$u \in U$	$w \in W$	edge weight $b(\{u, w\})$	
u_{C_j}	a	j	$j \in [l]$
u_{x_i}	a	l+i	$i \in [k]$
b_h	a	$h + \frac{1}{2}$	$h \in [k+l-1]$
u_{C_j}	$l_{1j}/l_{2j}/l_{3j}$	k+l+1	$j \in [l]$
u_{x_i}	x_i/\bar{x}_i	k+l+1	$i \in [k]$
u_{C_j}	v_{x_i}	(k+l+1) + (k+l)j + i	$i \in [k], j \in [l]$
u_{x_i}	$v_{x_{i'}}$	(k+l+1) + (k+l)(l+i) + i'	$i, i' \in [k], i' \le i$
u_{C_j}	$v_{C_{i'}}$	(k+l+1) + (k+l)j + k + j'	$j, j' \in [l], j' \leq j$

Table 2: Preferences of agents in the instance for Theorem 2. For correlated preferences, we again specify the edge weights of a significant subset of matching edges, which are used to derive the preference lists. For each edge not given here, there is a very small weight, and these edges are then ranked in arbitrary order at the bottom the preference lists of the incident agents. They do not change the dynamics as described in the proof.

Hence, the only way for such an agent u_s to become matched is via resolution of local blocking pairs that evolve when the left neighbor b_h is matched (cf. Fig. 3). Furthermore, with the exception of b_1 , none of the b_h agents is at hop distance 2 from any agent of W in the empty matching. Hence, in a similar fashion, such an agent can only get matched via a local blocking pair that evolves when the corresponding left neighbor $u_{s'}$ is matched. The latter local blocking pairs, however, can evolve only when u_s is matched to a, since all other agents of W rank all agents b_h at the bottom of their lists. As a consequence, the only way to create a matching edge for an agent b_h is by matching a first to a_{c_1} , then to a_{c_1} , to a_{c_2} , then to a_{c_1} , to a_{c_2} , etc. Note that once an agent a_{c_2} is matched to a_{c_3} , there is no local blocking pair involving a_{c_3} , since a_{c_4} is the most preferred partner for every a_{c_4} . Hence, to have the sequence proceed towards matching the a_{c_4} -agents, we must match a_{c_4} -a some agent a_{c_4} -and a_{c_4} -agents in the branching appear.

Now suppose some u_{x_i} is matched to a, and we resolve a local blocking pair with an agent in the branching. Suppose $u_{x_{i-1}}$ is unmatched at this point, and suppose all $u_{x_{i'}}$ with i' < i-1 are matched. If u_{x_i} does not become unmatched, there will be no local blocking pair involving a and $u_{x_{i-1}}$. Thereby a cannot become matched to $u_{x_{i-1}}$, and the latter will remain unmatched. Otherwise, if u_{x_i} becomes unmatched, this can only happen when matched to some agent $v_{x_1}, \ldots, v_{x_{i'}}$, at the point where that agent becomes matched to some $u_{x_{i'}}$ with i' < i-1. Then, however, it is easy to see that a similar key property as in the previous proof is violated $-u_{x_{i-1}}$ will not be able to discover $v_{x_{i-1}}$, since for every agent v_s matched to $u_{x_{i'}}$ there is no local blocking pair involving $u_{x_{i-1}}$. As such, $u_{x_{i-1}}$ will get "stuck" being matched to some agent on a path between a and $v_{x_{i'}}$.

If we apply the above argument inductively, it reveals that we must resolve local blocking pairs with agents in the branching and u_{x_i} in increasing order of i. The same argument shows that we have to match all u_{x_1}, \ldots, u_{x_k} to an agent in the branching before matching any of u_{C_1}, \ldots, u_{C_l} to agents in the branching. Together with the key property, this implies the desired structure. We need to match the agents u_{x_i} in a way to leave unmatched at least one literal agent for each clause. Hence, a sequence to reach M implies a satisfying assignment for the 3SAT formula.

In addition, NP-hardness also holds for reaching an arbitrary locally stable matching from $M = \emptyset$ under a restriction on the link structure. The proof of the following theorem was kindly provided by an anonymous referee.

Theorem 3. It is NP-hard to decide PATH-TO-STABILITY from the initial matching $M = \emptyset$ to an arbitrary locally stable matching in a bipartite network game with links restricted to $L \subseteq (U \times W) \cup (W \times W)$.

Proof. We define a stable matching M to be complete if all men and women are matched in M. Then, SMTI is the problem of deciding whether a complete stable matching exists in an instance of the stable marriage problem where preference lists may be incomplete and may contain ties. This problem is NP-complete, even if each woman's list is strictly ordered and each man's list is either strictly ordered or is a tie of length 2 [32]. We will use a reduction from this restriction of SMTI to show that it is NP-hard to decide PATH-TO-STABILITY from the initial matching $M = \emptyset$ to an arbitrary locally stable matching in a bipartite network game.

Let I be an instance of the stated restriction of SMTI, where U is the set of men and W is the set of women. We form an instance J of PATH-To-STABILITY by letting the potential matching edges E be obtained from the graph underlying I. That is, $\{u, w\}$ forms an acceptable pair in I if and only if $\{u, w\} \in E$. Initially, let L = E, i.e., the links comprise all potential matching edges. Each woman $w \in W$ initially has the same preference list in J as in I. We create a circling gadget G_w and identify woman w with vertex A_w . That is, woman w appends the men $3_w \succ 1_w \succ 2_w$, in that order, to her preference list in J. Let $U' \subseteq U$ be the set of men in I whose preference list is a single tie. Each man $u \in U \setminus U'$ has the same preference list in J as in I. Now let $u \in U'$ and suppose that u is indifferent between w_1 and w_2 in I. Remove $\{u, w_1, \}$ and $\{u, w_2\}$ from L. Create two new women a_u and b_u in J. Add $\{u, a_u\}$ and $\{u, b_u\}$ to E, and add $\{a_u, w_1\}$, $\{a_u, w_2\}$, $\{a_u, b_u\}$ and $\{u, b_u\}$ to L. In J, man u ranks $w_1 \succ w_2 \succ a_u \succ b_u$ in that order. It may be verified that I has a complete stable matching if and only if J has a sequence to a locally stable matching from the empty initial matching. Suppose that I has a complete stable matching M. We will show a path to a locally stable matching $M' \in J$. Pick any man $u \in U \setminus U'$. Then we may simply add the edge $\{u, M(u)\}\$ to M', where M(u) denotes u's partner in M. This is possible since $\{u, M(u)\}\in E\cap L$. Now let $u \in U'$. Then we add $\{u, b_u\}$ to M', since $\{u, b_u\} \in E \cap L$. Then u moves to a_u in M', since $\{a_u, b_u\} \in L$. Next u moves to M(u) in M', since $\{M(u), a_u\} \in L$. All women in W are now matched in M'. Therefore we can now let all circling gadgets stabilise in M', so that the final matching M' is locally stable in J.

Conversely suppose that J has a sequence to a locally stable matching M' starting from $M = \emptyset$. Since the circling gadgets must stabilize, every woman in W is matched in M' to a man in U outside of the circling gadget. Thus $M = M' \cap (U \times W)$ is a complete stable matching in I.

The following corollary provides a slightly weaker result than Theorem 3. It proves hardness either under the condition $L \subseteq (U \times W) \cup (W \times W)$ or the condition $M = \emptyset$. We include it since the proof construction is useful below to show additional hardness results for quality memory.

Corollary 1. It is NP-hard to decide PATH-TO-STABILITY to an arbitrary locally stable matching in a bipartite network game with links restricted to $L \subseteq (U \times W) \cup (W \times W)$. The same result holds for the initial matching $M = \emptyset$.

Proof. We adapt the previous constructions of Theorem 1 and 2. Instead of setting a specific matching M to converge to, we make use of circling gadgets. For every variable and every clause we use one separate circling gadget. We identify agents v_{x_i} and v_{C_j} in our previous constructions with agent A of the corresponding circling gadget. In Table 3 we show how to adjust the preference of the main structure from Theorem 1, which shows the result starting from $M_0 = \emptyset$. The adjustment of the main structure of Theorem 2 is very similar. Note that in the latter case we get $U = \{u_s, 1_s, 2_s, 3_s, | s \in \{x_1, \ldots, x_k, C_1, \ldots, C_l\}\}$, and hence this shows the result for $L \subseteq (U \times W) \cup (W \times W)$.

Agent	Preference List	
u_{x_i}	$v_{x_i} \succ v_{x_{i-1}} \succ \ldots \succ v_{x_1} \succ x_i \succ \overline{x}_i \succ a \succ [\ldots]$	$i \in [k]$
u_{C_j}	$v_{C_j} \succ \ldots \succ v_{C_1} \succ v_{x_k} \succ \ldots \succ v_{x_1} \succ l_{1j} \succ l_{2j} \succ l_{3j} \succ a \succ [\ldots]$	$j \in [l]$
b_h	$a \succ [\ldots]$	$h \in [k+l-1]$
a	$u_{x_k} \succ b_{l+k-1} \succ u_{x_{k-1}} \succ \ldots \succ b_1 \succ u_{C_1} \succ [\ldots]$	
v_{x_i}	$u_{C_l} \succ \ldots \succ u_{C_1} \succ u_{x_k} \succ \ldots \succ u_{x_i} \succ 3_{C_j} \succ 1_{C_j} \succ 2_{C_j} \succ [\ldots]$	$i \in [k]$
v_{C_j}	$u_{C_l} \succ \ldots \succ u_{C_{j+1}} \succ u_{C_j} \succ 3_{x_i} \succ 1_{x_i} \succ 2_{x_i} \succ [\ldots]$	$j \in [l]$
x_i/\overline{x}_i	$u_{x_i} \succ u_{C_1} \succ u_{C_2} \succ \ldots \succ u_{C_l} \succ [\ldots]$	$i \in [k]$

Table 3: Preference lists for agents of the main structure of Corollary 1 when adding circling gadgets to the hardness instances of Theorem 1. [...] indicates that all other remaining agents from the other partition follow in arbitrary order. The preference list of a_1 can be given arbitrarily.

For correctness, it suffices to reason why matching edges $\{\{u_s,v_s\} \mid s \in \{x_1,\ldots,x_k,C_1,\ldots,C_l\}\}$ are required to exist in every reachable locally stable matching. Then correctness follows using the arguments of the previous proofs. Assume there exists a reachable locally stable matching with some v_s not matched to any of the u-agents. For this s the circling gadget will not stabilize. v_{C_l} must be matched to u_{C_l} in order to stabilize the circling gadget for C_l , which leaves only $u_{C_{l-1}}$ for $v_{C_{l-1}}$ etc. Inductively, all the edges $\{\{u_s,v_s\}\mid s\in \{x_1,\ldots,x_k,C_1,\ldots,C_l\}\}$ must be present in every reachable locally stable matching.

3.2 Length of Sequences

For classic two-sided stable matching without locality constraints, from every initial state a stable matching can be reached in a polynomial number of steps [38]. With the restriction to local blocking pairs, we know that locally stable matchings exist, but the circling gadget shows that there are initial states from which none of the locally stable matchings is reachable. A weaker condition one might hope for is that once a state (locally stable or not) is reachable, it can also be reached via a short path of polynomially many steps. We show in Theorem 4 below that this condition holds for instances with correlated preferences, even beyond bipartite games in the roommates case. For uncorrelated strict preferences, however, we provide in Corollary 2 a class of instances with initial states such that every path to stability takes an exponential number of steps. This raises the question as to whether Path-To-Stability is even in NP.

Theorem 4. For every network game with correlated preferences, every (locally stable) matching $M^* \subseteq E$ and initial matching $M_0 \subseteq E$ such that M^* can be reached from M_0 through local improvement steps, there exists a sequence of at most $O(|E|^3)$ local improvement steps leading from M_0 to M^* .

Proof. Consider an arbitrary sequence between M_0 and M^* . We will show that only a polynomial part of it is necessary by omitting a possibly large number of intermediate steps. We rank all edges by their benefit (allowing multiple edges to have the same rank) such that r(e) > r(e') iff b(e) > b(e'). Further, we set $r_{max} = \max\{r(e) \mid e \in E\}$. First, recall the structure of the path of length at most 2 for accessible pairs discussed in Section 2.1. As a consequence, every edge e formed at any point during the sequence can be assigned to have at most one direct predecessor e' in the sequence, which was necessary for e to become accessible and get formed. Since e and e' must share a common agent, e' must get removed once e is formed. Hence, we call e the direct successor of e'. Also, note that being a local blocking pair implies r(e') < r(e). Thus, every $e \in M^*$ has at most

 r_{max} predecessors and every $e \in M_0$ at most r_{max} successors. We will say that upon formation e removes its predecessor e'. Note that the formation of e might also lead to removal of a second matching edge, which is not (assigned as) the predecessor of e. The remaining proof is based on three crucial observations:

- (1) Consider an edge e that has no predecessor that is, either $e \in M_0$, or $e \notin M_0$ with both agents connected via a path of at most two links. For every such e there is a *chain of edges* resulting from direct successor/predecessor relations. We say e starts a chain.
- (2) Consider $e \notin M_0$ that starts a chain. If the final edge of the chain is not in M^* and none of the edge formations of the chain removed any non-predecessor edge, the chain does not contribute in any way to the outcome and can be omitted completely from the sequence.
- (3) An edge can only be removed when forming a more preferred one. Hence, every chain is limited in length by r_{max} .

If $M_0 = \emptyset$, every edge that is created in the sequence is part of a chain that starts with $e \notin M_0$. Now, by (inductive application of) (2) we can restrict to creation of the chains with final edges in M. Thus, there are $|M^*|$ chains, and by (3) each of them has length at most r_{max} . Thus, there is a sequence to M of length at most $|M^*| \cdot r_{max} \in O(|E|^2)$.

If M_0 is arbitrary, every edge in M_0 starts a chain. Consider such a chain C where the final edge is not in M. Then, the final edge of C is removed by an edge that part of exactly one other chain C'. By (2), this can justify the existence of at most one other chain C'. Note that the final edge of C' must have a strictly higher rank than the final edge of C. Hence, after repeating this argument at most r_{max} many times, we arrive at a chain with a final edge in M. Each of these r_{max} many chains might need r_{max} many steps to arrive at its final edge. Thus, each edge in M_0 justifies the existence of at most r_{max} many chains of length r_{max} each. For all other edges of M, which cannot be traced back to a chain starting from M_0 , one can apply the previous arguments for the empty matching. Thus, there is a sequence to M of length at most $|M_0| \cdot r_{max} \cdot r_{max} + |M^*| \cdot r_{max} \in O(|E|^3)$. \square

Theorem 5. There is a bipartite network game with general preferences such that a locally stable matching can be reached by a sequence of local improvement steps from the initial matching $M = \emptyset$, but every such sequence has length $2^{\Omega(|V|)}$.

Proof. We compose the instance of several intertwined gadgets. Our instance consists of $n = \Omega(|V|)$ entangled gadgets $1, \ldots, n$, each of constant size, where w.l.o.g. let n be an even number. For clarity we list and analyze them as separately as possible, see below. For a schematic overview of the network construction, see Fig. 4.

The bipartite partition of the agent set is as follows:

$$U = \{A_0^o, End_0^e\} \cup \{A_i, B_i, C_i, End1_i, End2_i \mid i \text{ odd}\} \cup \{S_i^A, S_i^C, T_i^B, T_i^C \mid i \text{ even}\}$$

$$\cup \{D_i, 1_i, E_i, 2_i, F_i \mid i \text{ even}\} ,$$

$$W = \{A_0^e, b_0, End_0^o\} \cup \{A_i, B_i, C_i, End1_i, End2_i \mid i \text{ even}\} \cup \{S_i^A, S_i^C, T_i^B, T_i^C \mid i \text{ odd}\}$$

$$\cup \{D_i, 1_i, E_i, 2_i, F_i \mid i \text{ odd}\} .$$

Further, every End-agent is the agent A of a circling gadget and ranks the agents inside the circling gadget lowest.

We first explain how to construct a particular sequence from $M_0 = \emptyset$ to a locally stable matching. Then we argue that this sequence cannot be significantly shortened.

Generation Gadget:

Agents:

 $A_0^o, A_0^e, b_0^o, End_0^o, End_0^e$ Links: $\{\{A_0^o, C_i\} \mid i \text{ odd}\}, \{\{A_0^e, C_i\} \mid i \text{ even}\},\$

 $\{\{b_0, D_i\} \mid i \in [n]\}, \{A_0^o, b_0\}, \{A_0^e, b_0\},\$ $\{F_1, End_0^o\}, \{F_2, End_0^e\}, \{A_0^o, End_0^o\},$ $\{A_0^e, End_0^e\}$

Matching Edges:

 $\{\{A_0^o, D_i\} \mid i \text{ odd}\}, \{\{A_0^e, D_i\} \mid i \text{ even}\},\$ $\{A_0^o,1_1\},\,\{A_0^o,E_1\},\{A_0^o,2_1\},\{A_0^o,F_1\},$ $\{A_0^e,1_2\},\,\{A_0^e,E_2\},\{A_0^e,2_2\},\{A_0^e,F_2\},$ $\{A_0^o, End_0^o\}, \{A_0^e, End_0^e\}$

Agant	Drofovon as List
Agent	Preference List
10	$End_0^o \succ F_1 \succ 2_1 \succ E_1 \succ 1_1$
A_0	$\succ D_1 \succ D_3 \succ \ldots \succ D_{n-1}$
Δe	$End_0^e \succ F_2 \succ 2_2 \succ E_2 \succ 1_2$
A_0	$\succ D_2 \succ D_4 \succ \ldots \succ D_n$
b_0	_
End_0^o	A_0^o
End_0^e	A_0^e

Rotation Gadget i = 1:

Agents:

 $A_1, B_1, C_1, D_1, E_1, F_1, I_1, I_1, End I_1, End I_1$ Links:

 ${A_1, B_1}, {D_1, E_1}, {E_1, F_1}, {F_1, D_1},$ ${D_1, 1_1}, {1_1, E_1}, {E_1, 2_1}, {2_1, F_1},$ ${A_1, End1_1}, {C_1, End2_1}$ Matching Edges:

 ${A_1, E_1}, {A_1, F_1}, {B_1, D_1}, {B_1, E_1},$

 $\{C_1, F_1\}, \{C_1, D_1\}, \{F_1, End1_1\}, \{D_1, End2_1\}$

Agent	Preference List
A_1	$F_1 \succ E_1$
B_1	$E_1 \succ D_1$
C_1	$D_1 \succ F_1$
	$End2_1 \succ B_1 \succ T_2^B$
D_1	$\succ F_2 \succ 2_2 \succ E_2 \succ 1_2 \succ D_2$
	$\succ S_2^C \succ C_1 \succ A_0^o$
E_i	$A_1 \succ B_1 \succ A_0^o$
	$End1_1 \succ C_1 \succ T_2^C$
$ F_1 $	$\succ F_2 \succ 2_2 \succ E_2 \succ 1_2 \succ D_2$
	$\succ S_2^A \succ A_1 \succ A_0^o$
$1_1, 2_1$	A_0^o
$End1_1$	F_1
$End2_1$	D_1

Rotation Gadget i = 2:

Agents, Links, and Matching Edges similar as in all following gadgets $i = 3, \ldots, n-1$

Agent	Preference List
A_2	$F_2 \succ E_2$
B_2	$E_2 \succ D_2$
C_2	$D_2 \succ F_2$
	$End2_2 \succ B_2 \succ T_3^B$
D_2	$\succ F_3 \succ 2_3 \succ E_3 \succ 1_3 \succ D_3$
	$\succ S_3^C \succ C_2 \succ D_1 \succ F_1 \succ A_0^e$
E_2	$A_2 \succ B_2 \succ D_1 \succ F_1 \succ A_0^e$
F_2	$End1_2 \succ C_2 \succ T_3^C$
1'2	$\succ F_3 \succ 2_3 \succ E_3 \succ 1_3 \succ D_3$
	$\succ S_3^A \succ A_2 \succ D_1 \succ F_1 \succ A_0^e$
$1_2, 2_2$	$D_1 \succ F_1 \succ A_0^e$
S_2^C, T_2^C	
S_2^A, T_2^B	D_1
$End1_2$	F_2
End2	D_{α}

 $|End2_2|D_2$

Rotation Gadget i = 3, ..., n-1:

Agents:

 $A_i, B_i, C_i, D_i, E_i, F_i, 1_i, 2_i, S_i^A, S_i^C,$ $T_i^B, T_i^C, End1_i, End2_i$

 ${A_i, B_i}, {D_i, E_i}, {E_i, F_i}, {F_i, D_i},$ ${D_i, 1_i}, {1_i, E_i}, {E_i, 2_i}, {2_i, F_i},$ $\{A_i, End1_i\}, \{C_i, End2_i\}, \{D_i, S_i^A\}, \{S_i^A, A_{i-1}\},$ $\{D_i, S_i^C\}, \{S_i^C, C_{i-1}\}, \{F_i, T_i^B\}, \{T_i^B, B_{i-1}\},$ $\{F_i, T_i^C\}, \{T_i^C, C_{i-1}\}$

Matching Edges:

 ${A_i, E_i}, {A_i, F_i}, {B_i, D_i}, {B_i, E_i},$ ${C_i, F_i}, {C_i, D_i}, {F_i, End1_i}, {D_i, End2_i},$ ${D_i, D_{i-1}}, {1_i, D_{i-1}}, {E_i, D_{i-1}}, {2_i, D_{i-1}},$ ${F_i, D_{i-1}}, {D_i, F_{i-1}}, {1_i, F_{i-1}},$ ${E_i, F_{i-1}}, {2_i, F_{i-1}}, {F_i, F_{i-1}}$

Agent	Preference List
A_i	$ F_i \succ E_i $
B_i	$E_i \succ D_i$
C_i	$D_i \succ F_i$
	$End2_i \succ B_i \succ T_{i+1}^B$
D_i	$\succ F_{i+1} \succ 2_{i+1} \succ E_{i+1} \succ 1_{i+1} \succ D_{i+1}$
	$\succ S_{i+1}^C \succ C_i \succ D_{i-1} \succ F_{i-1} \succ A_0^{o/e}$
E_i	$A_i \succ B_i \succ D_{i-1} \succ F_{i-1}$
F_i	$End1_i \succ C_i \succ T_{i+1}^C$
	$\succ F_{i+1} \succ 2_{i+1} \succ E_{i+1} \succ 1_{i+1} \succ D_{i+1}$
	$\succ S_{i+1}^A \succ A_i \succ D_{i-1} \succ F_{i-1}$
$1_i, 2_i$	$D_{i-1} \succ F_{i-1}$
S_i^C, T_i^C	
S_i^A, T_i^B	D_{i-1}
$End1_i$	$ F_i $
$End2_i$	D_i

Rotation Gadget i = n:

Agents:

 $A_n, B_n, C_n, D_n, E_n, F_n, 1_n, 2_n, S_n^A, S_n^C,$ $T_n^B, T_n^C, End1_n, End2_n$ Links: ${A_n, B_n}, {A_n, C_n}, {B_n, C_n}, {D_n, E_n},$

 ${E_n, F_n}, {F_n, D_n}, {D_n, 1_n}, {1_n, E_n},$ ${E_n, 2_n}, {2_n, F_n}, {A_n, End1_n}, {C_n, End2_n},$ $\{D_n, A_{n-1}\}, \{D_n, C_{n-1}\}, \{F_n, B_{n-1}\}, \{F_n, C_{n-1}\}$ Matching Edges: ${A_n, E_n}, {A_n, F_n}, {B_n, D_n}, {B_n, E_n},$

 $\{C_n, F_n\}, \{C_n, D_n\}, \{F_n, End1_n\}, \{D_n, End2_n\},$ ${D_n, D_{n-1}}, {1_n, D_{n-1}}, {E_n, D_{n-1}},$ $\{2_n, D_{n-1}\}, \{F_n, D_{n-1}\}, \{D_n, F_{n-1}\}, \{1_n, F_{n-1}\},$ ${E_n, F_{n-1}}, {2_n, F_{n-1}}, {F_n, F_{n-1}}$

Agent	Preference List
A_n	$ F_n \succ E_n $
B_n	$E_n \succ D_n$
C_n	$D_n \succ F_n$
D_n	$ End2_n \succ B_n \succ C_n \succ D_{n-1} \succ F_{n-1} \succ A_0^e$
E_n	$A_n \succ B_n \succ D_{n-1} \succ F_{n-1}$
F_n	$ End1_n \succ C_n \succ A_n \succ D_{n-1} \succ F_{n-1}$
	$D_{n-1} \succ F_{n-1}$
S_n^C, T_n^C	$ F_{n-1} $
S_n^A, T_n^B	D_{n-1}
$End1_n$	$ F_n $
$End2_n$	D_n

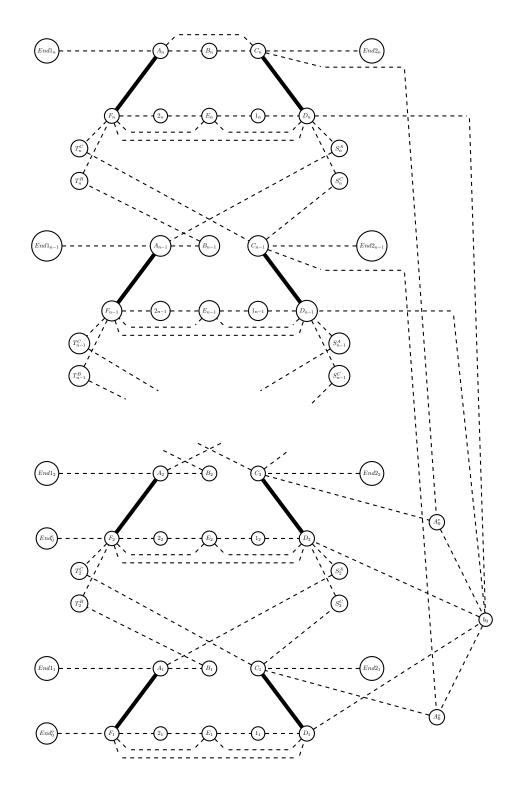


Figure 4: Network in the instance for Theorem 5. Matching edges depict a state in the described sequence. At this state, we have generated two edges in every rotation gadget, and the exponential rotation in the gadgets starts in order to iteratively get D_1 , E_1 and F_1 unmatched. This is necessary to match them to A_0^o , which thereby discovers End_0^o . Then, a similar approach allows A_0^e to get matched to End_0^e .

Obviously, since all agents $End1_i$, $End2_i$ are embedded into a circling gadget, we have to match them to D_i , F_i in order to achieve a locally stable matching. Hence, every locally stable matching must contain the edges $\{D_i, End2_i\}$ and $\{F_i, End1_i\}$. When these edges exist, we say rotation gadget i is fixed. For the same reason, edges $\{A_0^o, End_0^o\}$ and $\{A_0^e, End_0^e\}$ must exist in every locally stable matching. When these edges exist, we say the generation gadget is fixed.

The only local blocking pairs in $M = \emptyset$ are $\{A_0^o, D_i\}$ and $\{A_0^e, D_i\}$ for odd and even i, respectively. These agent pairs are accessible by a path of two links incident to b_0 . We start the sequence with creation of $\{A_0^o, D_1\}$. With the link $\{A_0^o, C_1\}$, the edge $\{C_1, D_1\}$ becomes accessible. At this position we could match D_1 to $End2_1$, thereby stabilizing the incident circling gadget. This, however, would match D_1 to its most preferred partner, and thereby we would not be able to construct $\{F_1, End1_1\}$. Hence, we instead match D_1 iteratively to agents of rotation gadget 2. We start by moving D_1 's edge from C_1 to S_2^C , and then further to D_2 , E_2 , E_2 , and then to E_2^D and E_2^D . Now we replace E_2^D and E_2^D and E_2^D , which is replaced by E_2^D and E_2^D are described by an arching iteratively to E_2^D and E_2^D and

Instead of fixing gadget 1, we proceed similarly with rotation gadget 2 by first creating $\{A_0^e, D_2\}$, then $\{C_2, D_2\}$ etc. until both $\{A_2, F_2\}$ and $\{C_2, D_2\}$ exist. Again, we will refrain from fixing rotation gadget 2, since this would imply that A_0^e cannot discover End_0^e , similar to the arguments before.

In a similar fashion we proceed with rotation gadgets $i=3,4,\ldots,n$, using A_0^o or A_0^e when i is odd or even, resp. Note that whenever we create edges in rotation gadget i, rotation gadget i+1 is still empty and all agents S_{i+1}^C , D_{i+1} , E_{i+1} , F_{i+1} , T_{i+1}^B are unmatched when D_i 's edge reaches them. At the end of this process, matching edges $\{C_i, D_i\}$ and $\{A_i, F_i\}$ exist for all $i=1,\ldots,n$. Observe that it is necessary to create two matching edges in each rotation gadget for the gadget to become fixed eventually. Since the only agents at which additional matching edges can be introduced are A_0^o and A_0^e , this has to be done before the generation gadget is fixed. This shows that every $\{C_i, D_i\}$ needs to be created before the generation gadget is fixed. We do not necessarily have to advance the sequence such that the first edge in each gadget reaches position $\{A_i, F_i\}$, but we have done so in our sequence for clarity and consistency.

Suppose a gadget is not fixed, then the two edges in this gadget can rotate along six positions. The *rotation* of a single edge runs as follows:

$$\{C_i, D_i\} \to \{B_i, D_i\} \to \{B_i, E_i\} \to \{A_i, E_i\} \to \{A_i, F_i\} \to \{C_i, F_i\} \to \{C_i, D_i\}$$

Note that since there are two edges, we always have four of the six agents matched. There is exactly one pair of agents unmatched, which allows to shift one of the edges further along its rotation. We call a sequence a *rotation of gadget* i if the two edges in gadget i both fulfill the complete cycle.

Only in gadget n, each of the six transitions is available directly via resolving a single local blocking pair. In all other gadgets $i=1,\ldots,n-1$, two transitions are not directly available as local blocking pairs. As we have seen above, for the transition from $\{C_i, D_i\}$ to $\{B_i, D_i\}$, D_i needs to discover B_i by matching iteratively to agents from gadget i+1. Exactly the same construction is in place for the transition $\{A_i, F_i\}$ to $\{C_i, F_i\}$, where the agents $S_{i+1}^A, D_{i+1}, E_{i+1}, F_{i+1}, T_{i+1}^C$ have to be iteratively matched to F_i for this agent to discover C_i . Since always only one of $D_{i+1}, E_{i+1}, F_{i+1}$, is unmatched, gadget i+1 also needs to complete a full rotation, thereby making $D_{i+1}, E_{i+1}, F_{i+1}$ iteratively available for the less preferred partners D_i or F_i . Consider the transition of, say, D_i from C_i to B_i . If D_i is matched to D_{i+1} , then in order to get E_{i+1} unmatched, we have to rotate the two edges in gadget i+1 to D_{i+1} and F_{i+1} . This, however, would imply we "lose" the edge incident to D_i , since D_i is currently still matched to D_{i+1} . Then, subsequently gadget i cannot be

fixed. Hence, in order to keep D_i matched, we match it to 1_{i+1} , proceed with the rotation in gadget i+1 to free E_{i+1} , then match D_i to E_{i+1} and further to 2_{i+1} . Finally, we complete the rotation of gadget i+1 by freeing F_{i+1} , which allows D_i to match to F_{i+1} , T_{i+1}^B and B_i . See Fig. 5 for a visualization of a complete rotation in gadget n to allow D_{n-1} become matched to D_n , E_n , F_n .

Since one rotation of gadget i involves two transitions, which each require a full rotation in gadget i+1, a single rotation of gadget 1 requires 2^{n-1} rotations of gadget n. Now, the single rotation of gadget 1 is required, since we need to make agents D_1, E_1, F_1 iteratively available for matching to A_0^o in order to fix the generation gadget. In the end, we rotate all edges back to $\{C_i, D_i\}$ and $\{A_i, F_i\}$ and fix the rotation gadgets. In this way, there is an exponential sequence in the gadget from $M = \emptyset$ to a locally stable matching.

Let us now argue, why the above idea of intertwined rotation cannot be avoided. The main underlying reason is that A_0^o and A_0^e must be used for generation of additional matching edges. Only after a sufficient number of matching edges is present in each gadget, we can proceed to fix the generation gadget. As part of this, we must fulfill a rotation in gadgets 1 and 2, implying an exponential number of rotations in higher gadgets.

Obviously, we cannot fix the generation gadget before two matching edges in each rotation gadget have been created. Similarly, none of the rotation gadgets can be fixed before the generation gadget is fixed – suppose a gadget i is fixed before the generation gadget is fixed. At this point, neither D_i nor F_i are willing to match to D_{i-1} or F_{i-1} . Thus, the rotation in gadget i-1 breaks and the gadget stabilizes. The same is true for all rotation gadgets $i-2,\ldots,1$. Hence, the rotation of gadget 1 cannot be completed, so A_0^o is unable to reach End_0^o . For the same reason, we must not create three or more matching edges in a rotation gadget. This stabilizes the gadget and does not allow rotation, which is needed to eventually fix the generation gadget.

A more subtle point is the interplay of even and odd gadgets. Since we strive for bipartite instances, we cannot use a single agent A_0 to generate matching edges in all rotation gadgets, since there are D_i -agents in both partitions. Given the sequence above, the reader might wonder why A_0^e has to trigger a final rotation in gadget 2. Imagine that A_0^e can directly match to End_0^e once all matching edges in the even rotation gadgets are created. Then, however, we could break the intertwining of rotation gadgets: First, create all edges in odd gadgets, and move A_0^o to End_0^o . Since at this point gadget 2 is empty, the single rotation in gadget 1 does not propagate to other gadgets. Then, create the matching edges of the even gadgets for gadget i = 2, 4, ..., n. This causes rotation in the odd gadget i+1, but since gadget i+2 does not yet contain any matching edges, the rotations do not propagate to higher gadgets. In this way, we could fix all gadgets in a polynomial number of steps.

In principle, we can also adopt this approach by first creating the matching edges in odd gadgets and then in even gadgets. However, in our instance A_0^e then triggers a final rotation in gadget 2 after generation of all matching edges in the even gadgets. At this point, all the gadgets are intertwined, and the single rotation in gadget 2 then requires that every gadget $i \geq 2$ rotates 2^{i-2} times.

4 Memory

In this section, we focus on the impact of memory for the reachability of locally stable matchings. As a direct initial result, we observe that no memory can help with the reachability of a *given* locally stable matching, even in a correlated job-market game.

Corollary 2. It is NP-hard to decide PATH-TO-STABILITY to a given locally stable matching in a correlated job-market game with any kind of memory.

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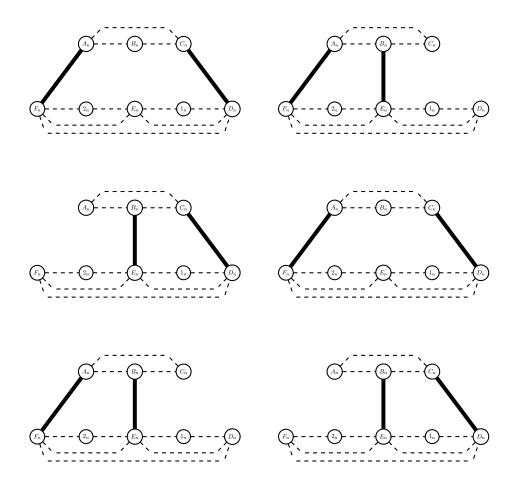


Figure 5: A rotation of gadget n. Suppose D_{n-1} must become iteratively matched to D_n, E_n, F_n to discover T_{n-1}^B and B_{n-1} . Top left: Starting position, D_n is occupied and should be freed. Top right: After resolution of $\{B_n, D_n\}$ and $\{B_n, E_n\}$, D_n is free to be matched to D_{n-1} . Middle left: By resolution of $\{C_n, F_n\}$ and $\{C_n, D_n\}$, D_n becomes matched, so D_{n-1} should become matched to D_n before this step. Middle right: After resolution of $\{A_n, E_n\}$ and $\{A_n, F_n\}$, E_n is free to be matched to D_{n-1} . Bottom left: By resolution of $\{B_n, D_n\}$ and $\{B_n, E_n\}$, E_n becomes matched, so D_{n-1} should get matched to D_n before this step. Bottom right: After resolution of $\{C_n, F_n\}$ and $\{C_n, D_n\}$, F_n is free to be matched to D_{n-1} , which subsequently gets matched to T_{n-1}^B . By resolution of $\{A_n, E_n\}$ and $\{A_n, E_n\}$ and $\{A_n, F_n\}$, the rotation is complete.

Proof. We observe that the same reduction as in Theorem 1 yields the result even for arbitrary memory. If the 3SAT formula is satisfiable, the sequence given in the proof can obviously be constructed without memory. Now assume that the desired matching M^* is reachable. Recall the instance, for which links and initial matching are depicted in Fig. 2, and the main idea of the proof. The agents of U iteratively get matched along the path of agents a_s , into the branching of agents x_i and \bar{x}_i , and subsequently to agents v_s on the right side of the path. For every agent $u \in U$, the preference of the partner is strictly increasing during this process. Hence, for u there is no (local) blocking pair with any agent $w \in W$ that u was matched to before. Thus, memory can only have an impact if an agent u gets unmatched.

An agent $u \in U$ might indeed become unmatched. For example, suppose we follow the sequence outlined in the proof of Theorem 1 by matching all agents u_{x_i} to an agent x_i or \bar{x}_i in the branching. Then, u_{C_1} gets matched iteratively to the agents along the path, to a, to an agent corresponding to a true literal, and then to v_{C_1} . At this point, we could resolve the local blocking pair of, say, u_{x_1} and v_{x_1} , thereby leaving u_{C_1} unmatched. u_{C_1} remembers all previous partners, so it forms blocking pairs, in particular, with all currently unmatched previous partners. In this way, u_{C_1} can get matched again, which would be impossible without memory. However, it is clear to see in this example that once v_{x_1} and u_{x_1} are matched, for all $j = 1, \ldots, m$, none of the edges $\{u_{C_j}, v_{C_j}\}$ can be constructed – none of them existed before, so v_{C_j} would have to be discovered by u_{C_j} via sequences of local blocking pair resolutions starting from v_{x_1} . However, $\{u_{x_1}, v_{x_1}\}$ is the top choice for both agents, and therefore none of them will become part of any local blocking pair again.

It is easy to verify that this condition must hold more generally. Every agent v_{x_j} must become matched to all agents u_{C_1}, \ldots, u_{C_l} before it gets matched to any of the agents v_{x_k}, \ldots, v_{x_j} . The argument is quite similar to the one in Theorem 2. Consider an agent v_{x_i} that becomes matched to $u_{x_{i'}}$ for $i \leq i'$ before getting matched to some u_{C_j} . Then u_{C_j} does not hold v_{x_s} in its memory, for all $s \geq i'$, since these agents must be discovered by iteratively resolving local blocking pairs of u_{C_j} and $v_{x_1}, \ldots, v_{x_{i'-1}}$. First, suppose $u_{x_{i'}}$ subsequently does not get unmatched. Then u_{C_j} will not be able to discover v_{x_s} for any $s \geq i'$, since no agent v is involved in a local blocking pair with u_{C_j} when being matched to $u_{x_{i'}}$. Also, none of these v_{x_s} can be made accessible for u_{C_j} by memory. Thus, figuratively speaking, the edge incident to u_{C_j} gets "stuck behind the edge incident to $u_{x_{i'}}$ when it tries to traverse the path of v-agents from v_{x_1} to v_{C_j} ".

Second, suppose $u_{x_{i'}}$ subsequently gets unmatched. This can only happen when some edge $\{u_{x_{i'}}, v_{x_s}\}$ with $i \leq s \leq i' - 1$ exists, and is replaced by $\{u_{x_{i''}}, v_{x_s}\}$ for some $s \leq i'' < i'$, since these are the only agents of U that v_{x_s} prefers to $u_{x_{i''}}$. Then, however, we can repeat the above argument that u_{C_j} is unable to reach v_{C_j} , where $u_{x_{i''}}$ takes the role of $u_{x_{i'}}$ and v_{x_s} takes the role of v_{x_i} .

This implies, in particular, that v_{x_1} must be matched to all agents u_{C_j} before getting matched to any of the agents u_{x_i} . Note that agent u_{x_i} only considers deviating to agents x_i and \bar{x}_i from the branching. This directly implies that while being matched to the a-agents or the x/\bar{x} -agents, no agent u_{x_i} can become unmatched. Therefore, we again must obey the structure of the sequence outlined in Theorems 1, 2 before, and a satisfying assignment must exist.

We will now concentrate on the impact of memory on reaching an arbitrary locally stable matching. Although existence is guaranteed for bipartite instances, even quite simple structures like the circling gadget do not allow a locally stable matching to be reached through improvement dynamics from every initial matching.

Quality Memory We start our treatment with quality memory, where every agent remembers at every round the best matching partner he ever had before. While this seems quite a natural choice

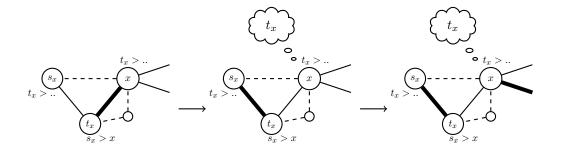


Figure 6: Effect of a memory jammer

and appears like a smart strategy for each agent, it can be easily fooled by starting with a muchliked partner who soon after matches with someone more preferred and never becomes unmatched again. This way the memory becomes useless which leaves us with the same dynamics as before.

Example (Memory Jammer). Consider the graph given in Fig. 6. As agents x and t_x are at distance 2 in the network (V, L), edge $\{x, t_x\}$ is accessible, regardless of the current matching. Furthermore, it is the only accessible edge for t_x when t_x is unmatched. Let t_x be the most preferred partner of x. Thus, $\{x, t_x\}$ is a local blocking pair as long as t_x is unmatched. Once this blocking pair gets resolved, edge $\{t_x, s_x\}$ becomes accessible. Let s_x be the most preferred partner of t_x , and t_x the most preferred partner of s_x . Thus, $\{t_x, s_x\}$ becomes a local blocking pair, and t_x will leave x to match to s_x . Since t_x is the most-liked partner, x will keep it in his memory, and t_x will never be replaced by any other agent. At the same time, t_x and t_x will never break their match regardless which other agents become accessible. Therefore, t_x has no benefit from remembering t_x , and all subsequent sequences will evolve in similar fashion as without memory of t_x .

We augment each agent in the circling gadget with a separate memory jammer. In the following Proposition 1, we show that this augmented circling gadget does not allow a sequence of local improvement steps from $M=\emptyset$ to any locally stable matching. We then show in Theorem 6 below that deciding this question is again NP-hard.

Proposition 1. There is a bipartite network game with general preferences, quality memory and initial matching $M = \emptyset$ such that no locally stable matching can be reached with local improvement steps from M.

Proof. As seen in the example above, we can add a memory jammer for every agent A, B, C, 1, 2, 3 of the circling gadget to make sure that initially each of the ordinary agents x gets to match with an agent t_x which symbolizes its top choice. Then t_x can (and will) switch to some agent s_x and stay in this matching for the rest of the dynamics. At the same time t_x will stay in the memory of gadget agent x indefinitely, thereby making his memory useless. Recall that we argued above, by suitably ranking the b_i -agents at the bottom of every list, they are never involved in any local blocking pair other than $\{i, b_i\}$. As such, we do not need memory jammers for b_i -agents.

Agent	Preference List
1	$t_1 \succ C \succ B \succ A$
2	$t_2 \succ A \succ C \succ B$
3	$t_3 \succ B \succ A \succ C$
A	$t_A \succ 3 \succ 1 \succ 2$
B	$t_B \succ 1 \succ 2 \succ 3$
C	$t_C \succ 2 \succ 3 \succ 1$
t_x	$s_x \succ x \text{ for } x \in \{1, 2, 3, A, B, C\}$
s_x	$t_x \text{ for } x \in \{1, 2, 3, A, B, C\}$

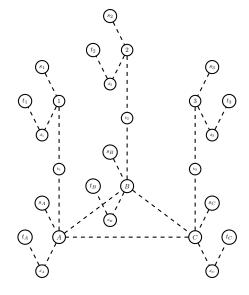


Figure 7: Preferences and links in the circling gadget with memory jammers as described in Proposition 1

Formally, we consider

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\begin{split} U = & \{1,2,3\} \cup \{t_x \mid x \in \{A,B,C\}\} \cup \{s_x \mid x \in \{1,2,3\}\} \\ W = & \{A,B,C,b_1,b_2,b_3\} \cup \{d_x \mid x \in \{1,2,3,A,B,C\}\} \\ & \cup \{t_x \mid x \in \{1,2,3\} \cup \{s_x \mid x \in \{A,B,C\}\} \} \\ L = & \{\{A,b_1\},\{B,b_2\},\{C,b_3\},\{b_1,1\},\{b_2,2\},\{b_3,3\}\} \\ & \cup \{\{A,B\},\{B,C\},\{C,A\}\} \\ & \cup \{\{s_x,x\},\{x,d_x\},\{d_x,t_x\} \mid x \in \{1,2,3,A,B,C\}\} \text{ and } \\ E = & \{\{u,v\} \mid u \in \{1,2,3\},v \in \{A,B,C\}\} \cup \{\{x,t_x\},\{t_x,s_x\} \mid x \in \{1,2,3,A,B,C\}\} \end{split}.
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The new preference lists and a sketch of the links can be found in Fig. 7. Note that we restrict the possible matching edges to the set of edges previously considered in the circling gadget and described for the memory jammers. It will become obviou below that by suitably ranking agents at the bottom of the lists, we can also assume that $E = U \times W$.

To show that no locally stable matching can be reached, start with $M=\emptyset$ and assume for contradiction that there is a sequence of local improvement steps leading to a locally stable matching. A matching cannot be locally stable until $\{x,t_x\}$ was created once for all $x\in\{1,2,3,A,B,C\}$, because otherwise it is accessible, matches x to his favorite partner, and matches t_x to the only possible accessible partner when being single. Afterwards, every such edge will be replaced by $\{t_x,s_x\}$, which remains stable since it matches both partners to their most preferred choice. Let x^* be the last agent for which $\{t_{x^*},s_{x^*}\}$ is generated. At that moment x^* is unmatched, and every agent of $\{1,2,3,A,B,C\}$ will continue to hold his t-partner in his memory. As the t-agents are not willing to change their matching edges, there will be no edge created from memory from this point on. If $x^* \in \{1,2,3\}$, one of the agents in $\{A,B,C\}$ is unmatched as well and vice versa. This leaves us in the situation described in the dynamics of the circling gadget. Since all memory entries are filled by t_x -agents, from this state no locally stable matching can be reached.

Remark. The same memory reset also works if every agent remembers the best k previous matches, for any number k, simply by applying k copies of the memory reset construction for each agent.

Theorem 6. It is NP-hard to decide PATH-TO-STABILITY to an arbitrary locally stable matching in a bipartite network game with quality memory.

Proof. For the proof we use the 3SAT gadget described in Corollary 2, where we proved that no memory can help to reach a given stable matching M. We then combine this instance with augmented circling gadgets (similar as we did in Corollary 1). In this way, we ensure that every stable matching must include the edges of M to stabilize the gadgets. This matching, however, can be reached if and only if the 3SAT formula is satisfiable.

As a slightly stronger result, the same hardness result might be shown with the empty initial matching. For this, one might be able to adjust the construction of Theorem 3 in order to avoid that quality memory helps reachability. We leave this issue as an avenue for future work.

Recency Memory In recency memory, every agent remembers the last agent it has been matched to before, which is yet another very natural update strategy. For our treatment we focus on the case in which the network links satisfy $L \subseteq (W \times W) \cup (U \times W)$. This is a natural restriction when one side consists of inanimate objects that form resources to be consumed. Theorem 3 shows NP-hardness of deciding Path-To-Stability in this class of instances. In contrast, with recency memory it is always possible to reach a locally stable matching.

Theorem 7. For every bipartite network game with general preferences, links $L \subseteq (U \times W) \cup (W \times W)$, recency memory and every initial matching, there is a sequence of $O(|U|^2|W|^2)$ many local improvement steps to a locally stable matching.

Proof. Our basic approach is to construct the sequence in two phases similarly as in [2]. In the first phase, we let the matched agents from U improve, but ignore the unmatched ones. In the second phase, we execute iterations, where in each iteration we pick a $u \in U$ and only do improvements steps involving u. Here we keep track of the agents from W and ensure they improve at the end of each iteration.

Preparation phase: As long as there is at least one $u \in U$ with u matched and u part of a local blocking pair, allow u to switch to the better partner.

The preparation phase terminates after at most $|U| \cdot |W|$ steps, as in every round one matched $u \in U$ strictly improves in terms of preference. This can happen at most |W| times for each matched u. In addition, the number of matched agents from U can only decrease.

Memory phase: As long as there is a $u \in U$ with u part of a local blocking pair, pick u and execute a sequence of local improvement steps involving u until u is not part of any local blocking pair anymore. For every edge $e = \{u', w\}$ with $u' \neq u$ that was deleted during the sequence, recreate e from the memory of u'.

We claim that if we start the memory phase after the preparation phase, at the end of every iteration we have the following invariants: The agents of W that have been matched before are still matched, they do not have a worse partner than before, and at least one of them is matched strictly better than before. Also, only unmatched agents from U are involved in local blocking pairs.

Obviously, at the end of the preparation phase, all local blocking pairs contain unmatched agents of U, i.e., initially our invariant holds. Let u be the agent chosen in the following iteration of the memory phase. At first, we consider the outcome for $w \in W$. If w is the agent matched to u in the

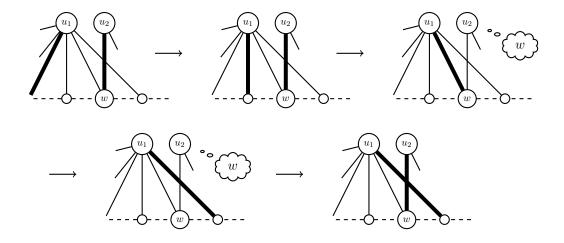


Figure 8: Example of a sequence computed by the algorithm in Theorem 7

end, then w clearly has improved. Otherwise w gets matched to its former partner (if it had one) through memory and thus is matched similarly as before. In particular, every w that represents an improvement to some u' but was blocked by a higher ranked agent, still remains blocked. Together with the fact that we execute local improvement steps involving u until it is not part of a local blocking pair anymore, this guarantees that all matched agents of U cannot improve at the end of this iteration of the memory phase. As one agent of W strictly improves in every iteration, we have at most $|U| \cdot |W|$ iterations in the memory phase, where every iteration consists of at most |W| steps by u and at most |U| - 1 edges reproduced from memory.

The structure of L, combined with recency memory, plays a key role in this proof. During the memory phase, an agent $u \in U$ can only become unmatched if its current partner from W deviates to a more desirable match. Due to the structure of L, agents in $U \setminus \{u\}$ and W do not require a matching edge to u in order to discover more desirable matches. Hence, it is not necessary to alter the memory of u to stabilize the system with respect to another agent $u' \in U$. Due to recency memory, agent u can reinsert the edge to the last partner, thereby allowing to recover the state of the system w.r.t. u and make overall progress towards a locally stable matching. We illustrate the arguments with the following example.

Example. In the example given in Figure 8 we execute an iteration of the memory phase. $u_1 \in U$ is the agent currently chosen. During the phase, u_1 repeatedly switches partner to improve. In that process, it also matches to agent w. w was matched to u_2 at the beginning of the iteration. However, w is only an intermediate partner for u_1 , who can improve further once new agents become accessible via the links of w. Thus, at the end of the iteration when u_1 is involved in no further local blocking pairs, the edge $\{u_2, w\}$ can be retrieved from u_2 's memory.

Remark. Note that the network used in Proposition 1 does not fulfill the criteria of $L \subseteq (U \times W) \cup (W \times W)$, and thereby the impossibility results for quality memory do not apply. However, if we discard all d-agents with their links as well as all $\{x, s_x\}$ -links, install direct links between t_x and s_x and use $\{\{x, t_x\} \mid x \in \{1, 2, 3\} \cup \{A, B, C\}\}$ as initial state, the memory gets jammed in the same way, and the social network now is limited to $L \subseteq (U \times W) \cup (W \times W)$. Theorem 6 easily continues to hold also with this type of circling gadget, which proves that for quality memory PATH-TO-STABILITY remains NP-hard even in instances with $L \subseteq (U \times W) \cup (W \times W)$.

The following is a direct corollary from the previous theorem. In random dynamics, with probability 1 in the limit, we will at least once start in a matching and execute the sequence described in the last theorem.

Corollary 3. For every bipartite network game with general preferences, links $L \subseteq (U \times W) \cup (W \times W)$, recency memory, and every initial matching, random dynamics converge to a locally stable matching in the limit with probability 1.

Sadly, we cannot always expect fast convergence here, as there are instances where random dynamics yield a sequence of exponential length with high probability even in ordinary stable marriage where all pairs of agents are accessible (see [2]).

Our constructions require only the agents from U to have recency memory, which allows us to show the same results when we have no memory for agents in W. In contrast, if we omit the memory for U and just allow recency memory for W, no sequence of local improvement steps might lead to a locally stable matching.

Proposition 2. There is a bipartite network game with general preferences, links $L \subseteq (U \times W) \cup (W \times W)$, recency memory for agents in W and initial matching $M = \emptyset$ such that no locally stable matching can be reached with local improvement steps from M.

Proof. Consider the circling gadget. First, as noted before, memory for the b_i -agents has no effect since the only matching edges that can evolve are $\{i, b_i\}$. All agents of W are matched in every locally stable matching, and none of the edges present in any locally stable matching is accessible via two links. As such, the last edge to complete any locally stable matching must come from memory. Now $\{1,A\}$, $\{2,B\}$, $\{3,C\}$ are always accessible. $\{1,C\}$, $\{2,A\}$, and $\{3,B\}$ only get removed if the W-agent finds a better partner. This shows that the locally stable matching $\{\{1,C\},\{2,A\},\{3,B\}\}$ cannot be reached, even with recency memory. By symmetry, assume the last edge recovered from memory be $\{2,A\}$. Then 2 must be the last partner A was matched to. Since A must leave 2 for a better partner (for 1), and $\{1,A\}$ must become deleted subsequently, A removes 2 from its memory.

As such, this leaves us with the locally stable matching $M^* = \{\{1, B\}, \{2, C\}, \{3, A\}\}\}$. Suppose the last edge of the matching created from memory is $\{3, A\}$. Then $\{2, B\}$ must have been created from the memory of B, since otherwise it needs $\{1, A\}$ to exist, which implies the memory of A is not set of A. For the same reason, A must have been created from memory, since for symmetric reasons A had not been able to be created from memory. Hence, in order to reach A the earliest step where the memory is set to A and remains set like this until the end. We say the memory for A is finalized in this step. Similarly, we consider the steps where the memories of agents A and A are finalized. Assume w.l.o.g. that the memory of A is finalized last. Then edge A must be constructed by existence of A. This, however, implies that the memory for A is finalized later, a contradiction.

The proposition can be used with the construction of Corollary 2 in the same manner as Proposition 1 was used to show Theorem 6. This yields the following result. As for quality memory, a similar result might be possible from the empty initial matching. We leave this issue as an open problem.

Theorem 8. It is NP-hard to decide PATH-TO-STABILITY to an arbitrary locally stable matching in a bipartite network game when recency memory exists only for one partition.

Random Memory Let us now consider how random memory might help us reach a locally stable matching from every starting state even in general bipartite network games. Again, we cannot expect fast convergence due to the full-information lower bound in [2]. However, we show that random memory can help with reachability:

Theorem 9. For every bipartite network game with random memory, random dynamics converge to a locally stable matching with probability 1.

Proof. Our proof combines an idea used in [19] with a convergence result in [38]. We consider a sequence of random local improvement steps and divide it into phases. Whenever a new edge is created for the first time, a new phase starts. Hence, phase t contains the part of the sequence, where exactly t edges have been created for the first time. Within a phase we have a fixed number of matching edges available (from the network or from memory). Consider phase t^* where t^* is the maximal phase of the sequence. t^* exists, as t is monotonically increasing and limited by |E|. We know that every phase $t < t^*$ ends after a finite number of steps. Hence we only have to show that phase t^* is finite in expectation. Roth and Vande Vate [38] demonstrate how to construct a sequence of blocking pair resolutions to form a stable matching with full information when all possible matching edges are accessible. In phase t^* , we have t^* edges that can be used for the matching, and all of them can also be remembered. Thus, with non-zero probability there is an initial state such that the random memory remembers the local blocking pair resolutions in the correct order and the random dynamics implement the local blocking pair resolutions in the correct way. Thus, phase t^* is ends after a finite number of steps in expectation. This proves the theorem.

5 Centralized problems

5.1 Maximum locally stable matching

The size of locally stable matchings in an instance can be highly variable – for example, in the trivial example where there are no links, the empty matching as well as a perfect matching are locally stable. From a designer's perspective, since all agents strive to be matched, it is desirable to form locally stable matchings of maximal size. Unfortunately, we will show that there is a close connection between maximum independent sets and maximum locally stable matchings. This allows us to transfer hardness of approximation results for independent set to locally stable matching.

Theorem 10. For every graph G we can build a network N such that N holds a maximum locally stable matching of size |V[G]| + k iff G has a maximum independent set of size k.

Proof. Consider a graph G = (V, E), |V| = n, as an instance of the maximum independent set problem. We construct the following job-market network $N = (V' = U \cup W, L)$. For every $v \in V$ we have agents $u_{v,1}$ and $u_{v,2}$ in U and agents $w_{v,1}$ and $w_{v,2}$ in W. Further there are links $\{w_{v,1}, w_{v',2}\}$ and $\{w_{v',2}, w_{v,2}\}$, if $v' \in N(v)$. We allow matching edges $\{u_{v,1}, w_{v,1}\}$, $\{u_{v,1}, w_{v',2}\}$ for $v' \in N(v)$, $\{u_{v,1}, w_{v,2}\}$ and $\{u_{v,2}, w_{v,2}\}$. Each $u_{v,1}$ prefers $w_{v,2}$ to every $w_{v',2}$, $v' \in N(v)$, and every $w_{v',2}$ to $w_{v,1}$. The preferences between the the different neighbors can be chosen arbitrarily. Each $w_{v,2}$ prefers $u_{v,1}$ to every $u_{v',1}$, $v' \in N(v)$, and every $u_{v',2}$ to $u_{v,2}$. Again the neighbors can be ordered arbitrarily. The agents $w_{v,1}$ and $u_{v,2}$ have only one possible matching partner anyway.

We claim that G has a maximum independent set of size k iff N has a locally stable matching of size n + k. First, let S be a maximum independent set in G. Then $M = \{\{u_{v,1}, w_{v,2}\} \mid v \in V \setminus S\} \cup \{\{u_{v,1}, w_{v,1}\}, \{u_{v,2}, w_{v,2}\} \mid v \in S\}$ is a locally stable matching in N. From edges $\{u_{v,1}, w_{v,2}\}$

no agent wants to deviate. For the other agents, the independent set property tells us that for $v \in S$ all agents $v' \in N(S)$ yield edges $\{u_{v',1}, w_{v',2}\}$ that keep $u_{v,1}$ from switching to $w_{v',2}$. Thus, from $\{u_{v,1}, w_{v,1}\}$ no agents wants to deviate. $w_{v,2}$ is not accessible with $u_{v,1}$, which implies that the agents do not want to deviate from $\{u_{v,2}, w_{v,2}\}$.

Now, let M be a maximum locally stable matching for N. Further, we chose M such that every $u_{v,1}$ is matched, which is possible as replacing a matching partner of $w_{v,2}$ by (the unmatched) $u_{v,1}$ will not generate local blocking pairs or lower the size of M. We note that no $u_{v,1}$ is matched to some $w_{v',2}$ with $v \neq v'$, since this would imply $\{u_{v,1}, w_{v,2}\}$ becomes accessible and a local blocking pair. Then for $S = \{v \mid u_{v,2} \in M\}, |S| = |M| - n$. Every $u_{v,2}$ can only be matched to $w_{v,2}$, which means that $u_{v,1}$ must be matched to $w_{v,1}$. These agents must not be involved in local blocking pairs, which implies for every $v \in S$, $N(v) \cap S = \emptyset$. Thus, S is an independent set.

Corollary 4. Under the unique games conjuncture maximum locally stable matching cannot be approximated within $1.5 - \epsilon$, where ϵ approaches 0 when n grows.

Proof. We combine the relation to maximum independent set as shown in Theorem 10 with a result of [7] that independent set is unique-games-hard to approximate within a factor of $\Omega\left(\frac{d}{\log^2(d)}\right)$ for independent sets of size $k = \left(\frac{1}{2} - \Theta\left(\frac{\log(\log(d))}{\log(d)}\right)\right)n$, where d is the maximum degree. Hence, it is hard to distinguish instances with maximum independent set of size k and $k \cdot O\left(\frac{\log^2(d)}{d}\right)$. With $d = \Theta(n)$, maximum locally stable matching is unique-games-hard to approximate within

$$\frac{n+k}{n+k\cdot O\left(\frac{\log^2(n)}{n}\right)} \ = \ \frac{n+\left(\frac{1}{2}-\Theta\left(\frac{\log(\log(n))}{\log(n)}\right)\right)n}{n+\left(\frac{1}{2}-\Theta\left(\frac{\log(\log(n))}{\log(n)}\right)\right)n\cdot O\left(\frac{\log^2(n)}{n}\right)} \ \le \ 1.5-\epsilon \ ,$$

for any constant $\epsilon > 0$.

Remark. Note that in fact we only used the setting of the bipartite job-market game, where one side has no network at all. This shows that even under quite strong restrictions the hardness of approximation holds.

Proposition 3. If a stable matching exists, every stable matching is a 2-approximation of a maximum locally stable matching.

Proof. Note that every stable matching is locally stable as well. Now let M be a stable matching and $e = \{u, v\}$ an edge of a maximum locally stable matching M^* . We show that at least one agent of e is matched in M. Then obviously $|M^*| < 2|M|$. Assume that both agents are unmatched in M. As e exists in M^* , u and v prefer each other to being alone. Thus $\{u, v\}$ is a blocking-pair and M cannot be stable.

5.2 Roommates Problem

If E is bipartite, there always exists a stable (and thus, a locally stable) matching. In the more general roommates case, there are instances with general preferences such that no stable matching exists. The same obviously holds for locally stable matchings, since locality has no effect if E contains a link for every pair of agents. While the existence of a stable matching can be decided in polynomial time [26], we show that the same question is NP-hard for locally stable matchings. Our initial proof was significantly simplified by an anonymous referee, who kindly provided the following proof.

Theorem 11. It is NP-complete to decide if a locally stable matching exists in a network game.

Proof. Let SRT denote the problem of deciding whether an instance of the stable roommates problem with ties admits a stable matching. SRT is NP-complete even if each agent finds all other agents acceptable, and each tie is of length 2 and occurs at the head of some agent's preference list [27].

We will use a reduction from this restricted version of SRT to show that our problem is NP-complete. Let I be an instance of the stated restriction of SRT, and V is the set of agents in I. Let $V' \subseteq V$ denote the set of agents in I who have a tie (of length 2) at the head of their preference list in I. We form an instance J of Local-SRT by letting the potential matching edges E be the edge set of the complete graph on V. Initially, let L = E, i.e., the links comprise all potential matching edges.

Each agent $v \in V$ initially has the same preference list in J as in I. If $v \in V'$, then assume that the tie at the head of v's preference list involves agents v' and v''. By inspection of the proofs of Theorem 6.1 and Corollary 6.3 in [27], for one of v' and v'' (w.l.o.g. let that be v'') the unique first partner on the preference list is v. Moreover, $v' \notin V'$ and $v'' \notin V'$. Break the tie in J so that v prefers v' to v'', and then v'' is preferred to all other agents on v's preference list, as before. Remove all links incident to v from L apart from the link $\{v, v''\}$. Moreover, remove the link $\{v', v''\}$ from L. Hence, now if v is unmatched or matched to v'', it is not accessible with v'.

We claim that I has a stable matching if and only if J has a locally stable matching. First, let M be a stable matching in I. Suppose that $\{v_1, v_2\}$ is a local blocking pair of M in J. If $v_1 \notin V'$ and $v_2 \notin V'$ then it is clear that $\{v_1, v_2\}$ is a blocking pair of M in I, a contradiction. Now suppose w.l.o.g. let $v_1 \in V'$. Let v_1' and v_1'' be the two members of v_1 's tie in I, and w.l.o.g. suppose v_1 prefers v_1' to v_1'' in J. By inspection of the proofs of Theorem 6.1 and Corollary 6.3 in [27], it may be verified that $M(v_1) \in \{v_1', v_1''\}$. It follows that $v_2 = v_1'$ and $M(v_1) = v_1''$. But by construction of L, v_1 and v_1' are not accessible, a contradiction. Hence M is locally stable in J.

Conversely, suppose that M is a locally stable matching in J. Suppose that $\{v_1, v_2\}$ is a blocking pair of M in I. If $\{v_1, v_2\} \in L$ then $\{v_1, v_2\}$ is a local blocking pair of M in J, a contradiction. Next suppose that $\{v_1, v_2\} \notin L$. If $v_1 \notin V'$ and $v_2 \notin V'$, then v_1 and v_2 are the two members of the tie of length 2 in some agent v's preference list. W.l.o.g. suppose that v prefers v_1 to v_2 in J. Note $\{v, v_2\} \in L$ and v_2 ranks v in first place, so $M(v) \in \{v_1, v_2\}$, for otherwise $\{v, v_2\}$ is a local blocking pair of M, a contradiction. By inspection of the proofs of Theorem 6.1 and Corollary 6.3 in [27], if $\{v, v_1\} \in M$, then v_1 prefers v to v_2 in I, whilst if $\{v, v_2\} \in M$, v_2 prefers v to v_1 in I. Hence $\{v_1, v_2\}$ cannot be a blocking pair of M in I after all. Finally, suppose $\{v_1, v_2\} \notin L$ and w.l.o.g. $v_1 \in V'$ and $v_2 \notin V'$. Let v_1' and v_1'' be the two members of v_1 's tie in I. If $M(v_1) \in \{v_1', v_1''\}$, then v_1 cannot be involved in any blocking pair in I. Otherwise, v_1 prefers each of v_1' and v_1'' to $M(v_1)$ in I. By construction of I, $\{v_1, v_1''\} \in L$ and v_1'' ranks v_1 in first place. Hence $\{v_1, v_1''\}$ is a local blocking pair of M in I, a contradiction. Thus M is stable in I.

6 Conclusion

In this paper, we study PATH-TO-STABILITY problems for locally stable matchings. Our results show that locally stable matchings might not necessarily be reachable by a sequence of local improvement steps. We prove that deciding this question is NP-hard in many rather restricted domains of two-sided instances. Moreover, even in cases where a locally stable matching can be reached from an initial matching, the length of the shortest such sequence can be exponentially long. An interesting case, where locally stable matchings can always be reached, are instances with correlated preferences [19]. For these instances we show that, in fact, every reachable matching can be reached

within $O(|E^3|)$ many steps. Another interesting and natural property, which can be used to overcome these negative conditions, is memory. With recency memory we can guarantee existence of a sequence of $O(|U|^2|W|^2)$ many steps to a locally stable matching when there is no network among one partition. With random memory convergence is guaranteed in the limit with probability 1.

In this direction, there are a variety of open problems. It is not known if the polynomial bounds we provide for correlated preferences or recency memory are tight. Furthermore, it would be interesting to see if there are other meaningful classes of instances, for which existence of a path to stability (of polynomial length) can be shown, or at least decided in polynomial time.

In addition to convergence questions, we study existence and optimization problems regarding locally stable matchings. For finding the locally stable matching of maximum cardinality, we provide a lower bound of $1.5 - \epsilon$ on the approximation factor under the unique games conjecture, for every constant ϵ . In the roommates case, even existence of locally stable matchings is shown to be NP-complete to decide.

Perhaps the most interesting open problem in this domain is whether there is an efficient 1.5-approximation algorithm for finding a maximum locally stable matching. Alternatively, can the lower bound can be strengthened towards the simple upper bound of 2? For the roommates case, it would be interesting to identify further classes of instances, for which existence of a locally stable matching is guaranteed or at least decidable in polynomial time.

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