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Abstract This paper considers marriage problems, roommate problems with nonempty core, and college admissions problems with responsive preferences. All stochastically stable matchings are shown to be contained in the set of matchings which are most robust to one-shot deviation.

Keywords: Learning; stochastic stability; matching; marriage; college admissions.

JEL Classification Numbers: C71, C72, C73, C78, D71.

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1 Introduction

Partnerships fail. Marriages break down, friendships rupture, your gym buddy stops training. When partnerships break down, new partnerships are forged in the aftermath, until an equilibrium, or something close to an equilibrium, is again reached. The reasons that partnerships can break down are many: often human imperfection and the vicissitudes of fate play a role. Errors, mishaps or misbehavior on the part of one of the partners can contribute to the decline of a partnership. However, partnerships are not all alike. Some partnerships are strong, some are weak. Some partnerships are easily substitutable, others less so. It is not just the partnerships themselves that can be more or less robust. Due to the interrelationship of different partnerships, networks of partnerships also display robustness characteristics which depend on the robustness of their constituent pairings. This paper analyses such settings in the context of the well known marriage problem of Gale and Shapley (1962) as described in Jackson and Watts (2002). We show that for standard matching dynamics, perturbed by an error process, any stochastically stable matching is contained in the class of matchings which are most robust to one-shot deviation. For the logit choice rule, this class corresponds to a non-transferable utility version of the least-core as described in Maschler, Peleg and Shapley (1979). The results extend to one-sided matching markets (roommate problems) and to many-to-one matchings (college admissions problems).

Similarly to the related papers of Jackson and Watts (2002) and Klaus, Klijn and Walzl (2010), players occasionally make mistakes in a dynamic model of partnership formation. Mistakes involve a player leaving an existing partner or matching with a new partner in such a way that his payoff is reduced. Mistakes can be fatal to a partnership and can drive the dynamic process of partnership formation to a new equilibrium. Jackson and Watts (2002) and Klaus et al. (2010) derive results for stochastic stability of marriage and roommate problems under uniform mistake probabilities: every possible mistake has the same order of probability of occurring. They find that all stable matchings are stochastically stable in the sense of Kandori, Mailath and Rob (1993); Young (1993).\footnote{In the language of Noldeke and Samuelson (1993), they find that all the equilibria are part of a single mutation connected component. This fact means that all of the stable partnership networks in their setting are stochastically stable under uniform mistakes.} When mistakes are rare, in the long run the process will spend almost all of its time at stochastically stable matchings.

The current paper addresses a large class of alternative mistake models, including payoff dependent models such as logit choice (Blume, 1993) and probit choice (Dokumaci and Sandholm, 2011; Myatt and Wallace, 2003), for which the decision rule depends ex-
plicitly on cardinal preferences. Due to differing strengths of partnerships, the authors of the current paper believe cardinal preferences to be a natural assumption in matching models. Moreover, abstraction away from cardinal preferences, or the choice of a dynamic which is insensitive to such preferences, is not without loss when it comes to applying a concept such as stochastic stability. As pointed out by Bergin and Lipman (1996), the identity of stochastically stable states depends on the mistake model. Therefore, results which are applicable across a broad range of mistake models are of particular interest.

In this paper, it is no longer the case that all stable matchings are stochastically stable. Given the large class of mistake models considered, this is unremarkable (Bergin and Lipman, 1996). What is remarkable is that for all of these models there exists a simple local property that must be satisfied by any stochastically stable matching. Specifically, this paper shows that stochastically stable matchings must be contained in the set of stable matchings which are most robust to one-shot deviation, that is the set of stable matchings at which the most probable mistake is not more probable than the most probable mistake at any other stable matching. When this set is a singleton, as at least one stochastically stable state always exists, its unique member must be the unique stochastically stable state. These results hold for marriage problems, roommate problems with nonempty core, and college admissions problems with responsive preferences.

Our result is surprising because stochastic stability is a globally determined property: existing characterizations (Kandori et al., 1993; Young, 1993) and partial characterizations (Ellison, 2000) depend on transition paths between all of the stable states. Computing probabilities for all such transition paths can be cumbersome. In contrast, the set of stable matchings which are most robust to one-shot deviation is defined solely by reference to local properties of the stable matchings. To compare with another contribution to the partial characterization literature, Ellison (2000) provided a globally determined sufficient condition for stochastic stability in any (finite) problem, whereas we provide a locally determined necessary condition for stochastic stability in matching problems.

For the logit choice rule, the most probable mistake at a stable matching is the deviation which causes the lowest payoff loss to the deviating players. The set of stable matchings which are most robust to one-shot deviation maximize this lowest possible payoff.

\footnote{The relation of such rules to uniform mistake models can be thought of as similar to the relation between the static concepts of Proper Equilibrium (Myerson, 1978) and Trembling Hand Perfect Equilibrium (Selten, 1975). In the former, mistakes associated with larger payoff losses are less likely, whereas in the latter there is no difference.}

\footnote{It is not uncommon to assume cardinal preferences in the literature on matching problems. Abdulkadir-Diroğlu, Che and Yasuda (2011) discuss that a mechanism sensitive to cardinal preferences may achieve a Pareto-superior matching to one obtained by the deferred acceptance mechanism of Gale and Shapley (1962).}
loss. This set is equal to a non-transferable utility version of the least-core as described in Maschler et al. (1979). That is, there is a connection between perturbed adaptive dynamics, matching problems, robustness to one-shot deviations and a well known concept in cooperative game theory.

There is a growing literature which looks at equilibrium selection in matching problems (Biró and Norman, 2013; Boudreau, 2012; Echenique and Yariv, 2012; Pais, Pinter and Veszteg, 2012). Typically, these papers use simulation or experimental evidence to generate a distribution over absorbing states reached by a dynamic process without mistakes, conditional on the process being started at some initial matching. In contrast to these papers, our results are independent of the initial matching and the probabilities with which any players are chosen to better respond. Moreover, the results in the current paper are analytical. The papers cited above consider short run behavior given some initial condition. In contrast, the current paper models the long run.

A useful literature from the perspective of the current paper is the paths to stability literature in matching problems with non-transferable utility. This focuses on convergence to core allocations in situations where the payoff for an individual depends only on his partner (Diamantoudi, Xue and Miyagawa, 2004; Roth and Vande Vate, 1990). Another related literature is the literature on convergence to the core in cooperative games (Agastya, 1997; Feldman, 1974; Green, 1974; Newton, 2012). A branch of this literature has recently explicitly focused on the case in which all relevant coalitions are pairs – the transferable utility equivalent of the marriage problem, otherwise known as the assignment problem (Biró, Bomhoff, Golovach, Kern and Paulusma, 2012; Chen, Fujishige and Yang, 2010; Nax and Pradelski, 2013; Shapley and Shubik, 1971). Of particular note is the work of Nax and Pradelski (2013), who adapt the results of the current paper to obtain selection within the core of the assignment game.

The paper is organized as follows. Section 2 gives the model and some relevant concepts from the literature. Section 3 gives the main results for marriage problems. Section 4 applies our results to marriage problems under differing choice rules. Sections 5 and 6 extend our main result to many-to-one matching problems and to roommate problems respectively.

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4Boudreau (2011) writes of the prior approach: “Calculating the probability of each stable outcome for a given market under the randomized tâtonnement process is extremely difficult due to the tremendous number of paths that can be involved. . . Loops in the process mean that a closed form solution is virtually impossible to obtain.”
2 Model

2.1 The marriage problem

We follow the description of the marriage problem in Jackson and Watts (2002). There is a set of players, $N$, which is divided into a set of men, $M = \{m_1, \ldots, m_k\}$, and a set of women, $W = \{w_1, \ldots, w_l\}$. An undirected network $g$ is a set of edges $ij \in g$, each comprising a pair of players $i, j \in N$, $i \neq j$, such that $ij \in g \iff ji \in g$. Let $G$ denote the set of all undirected networks on $N$. Let $g(i) = \{j : ij \in g\}$ denote the set of players linked to player $i$ in network $g$. $g(i) = \emptyset$ means that $i$ is single in $g$. The set of matchings in the marriage problem, $G$, is the set of undirected networks in which each woman is linked to at most one man, and each man is linked to at most one woman:

$$G = \{g \in G : (\forall ij \in g, i \in M \iff j \in W), (\forall i \in N, |g(i)| \leq 1)\}.$$

In a slight abuse of notation, we sometimes write $g(i) = j$ for $g(i) = \{j\}$. Let $\mu = \{(i, j) : (\exists g \in G : ij \in g)\}$ be the set of pairs of players between whom a link can potentially exist.

The vector of utilities obtained from network $g$ by the players is given by $u : G \to \mathbb{R}^{|N|}$. Player $i$ obtains utility $u_i(g)$ from network $g$, and this utility depends only on the match of $i$. That is, for each $i$, $u_i(g) = u_i(g')$ if $g(i) = g'(i)$. We assume that players are never indifferent between two potential matches: $g(i) \neq g'(i)$ implies that $u_i(g) \neq u_i(g')$. Therefore, if $g(i) \neq \emptyset$, then $u_i(g) = u_i(\{ig(i)\})$, and if $g(i) = \emptyset$, then $u_i(g) = u_i(\emptyset)$. Define $g - ij := g \setminus \{ij\}$ as the network $g$ with the edge $ij$ removed if it exists in $g$. Similarly, define $g + ij := (g \setminus \{kl : k = i, l \in g(i) \text{ or } k = j, l \in g(j)\}) \cup \{ij\}$ as the network $g$ with the edge $ij$ added and any existing edges exiting $i$ and $j$ removed.

**Definition 2.1** A matching $g \in G$ is stable if:

(i) $\forall ij \in g, \ u_i(g) > u_i(g - ij)$.

(ii) $\exists i \in M, j \in W : \ u_i(g + ij) > u_i(g)$ and $u_j(g + ij) > u_j(g)$.

We denote the set of stable matchings by $\mathcal{C}$. The set of stable matchings corresponds to the core of the problem: the set of matchings from which no subset of players can improve their payoffs by removing and adding edges in a coordinated manner.
2.2 Unperturbed blocking dynamic

We describe a class of unperturbed blocking dynamics. Let $g^t$ be the network in period $t$. At the beginning of period $t + 1$, a pair of players $(i, j)$ is selected at random according to a distribution $F_{g^t}(.)$ with full support on $\mu$. Let $g^{t+1}$ be determined as follows:

(i) If $g^t(i) = j$ and either $u_i(g^t - ij) > u_i(g^t)$ or $u_j(g^t - ij) > u_j(g^t)$, then, with some probability greater than zero, set $g^{t+1} = g^t - ij$.

(ii) If $g^t(i) \neq j$, $u_i(g^t + ij) > u_i(g^t)$ and $u_j(g^t + ij) > u_j(g^t)$ then, with some probability greater than zero, set $g^{t+1} = g^t + ij$.

(iii) $g^{t+1} = g^t$ otherwise.

In the terminology of matching problems, a pair $(i, j) \in \mu$ blocks a matching $g$ if they prefer one another to their partners in $g$. Denote the transition probabilities of a given unperturbed blocking dynamic by $P_0(\ldots)$. That is, $P_0(g, g')$ is the probability that $g^{t+1} = g'$, given that $g^t = g$.

2.3 Perturbed blocking dynamic

Players meet and will usually take the myopically optimal action, whether that is to stay with their current partner, dissolve an existing partnership, or create a new partnership. However, from time to time, players make mistakes and take actions which reduce their payoffs, whether it be leaving or creating a partnership. That is, a pair selected by the dynamic will sever an existing beneficial link, or create a link which is worse than the status quo for at least one of the players involved. We consider families of perturbed blocking dynamics, with transition probabilities $P_\eta(\ldots)$, indexed by a parameter $\eta \in (0, \bar{\eta})$.

The family $\{P_\eta\}_{\eta \in (0, \bar{\eta})}$ is assumed to satisfy the following conditions.

Assumption 1 (conditions on the perturbed dynamic)

(i) $P_\eta \xrightarrow{\eta \rightarrow 0} P_0$, where $P_0$ are the transition probabilities for some unperturbed blocking dynamic as described in Section 2.2.

(ii) For $\eta > 0$, the chain induced by $P_\eta$ is irreducible.

(iii) $P_\eta$ vary continuously in $\eta$.

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5Our unperturbed dynamic is essentially the same as those of Roth and Vande Vate (1990), Jackson and Watts (2002) and Klaus et al. (2010).
(iv) If, for \( g \neq g' \), \( P_0(g, g') = 0 \), \( P_\hat{\eta}(g, g') > 0 \) for some \( \hat{\eta} > 0 \), then \( \lim_{\eta \to 0} -\eta \log P_\eta(g, g') = c \) for some \( c > 0 \).

(v) For any \( \eta \geq 0 \), \( P_\eta(g, g') > 0 \) implies \( g' = g + ij \) or \( g' = g - ij \) for some \( (i, j) \in \mu \).

Condition (i) merely states that the family of perturbed dynamics corresponds to an unperturbed dynamic. Conditions (ii), (iii), (iv) restrict the process to weakly regular Markov chains. A broad class of strategy revision rules falls into this category. Examples include best response with mutations, the logit choice rule, pairwise comparison rules, and the probit choice rule (see Sandholm, 2010). Condition (v) means that transitions always involve a single pair of players getting together or splitting up. This restriction is needed to eliminate the possibility of two couples separating (or getting together) at the same point in time with a higher (order of) probability than either one of the couples acting alone.

As a chain with \( \eta > 0 \) is irreducible, there exists a unique stationary distribution \( \pi_\eta \). For convenience, we assume the following.\(^6\)

**Assumption 2 (existence of limit)**

\[
\pi_0 := \lim_{\eta \to 0} \pi_\eta \quad \text{exists.}
\]

A matching \( g \) is stochastically stable if \( \pi_0(g) > 0 \). We denote the set of stochastically stable states by \( SS \).

**Definition 2.2**

\[
SS := \{ g \in G : \pi_0(g) > 0 \}
\]

All stochastically stable matchings belong to recurrent classes of the unperturbed process (Young, 1993) and from any matching there exists a finite sequence of transitions under the unperturbed process that culminates in a stable matching being reached (Jackson and Watts, 2002; Roth and Vande Vate, 1990). Therefore, the only recurrent classes of the unperturbed process are the individual stable states, i.e. \( SS \subseteq \mathcal{C} \). The identity of the stochastically stable matchings is important, as for small error probabilities the process will spend almost all of the time at these matchings.

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\(^6\)This condition could be avoided if stochastically stable states were defined as states for which \( \pi_\eta(.) \not\to 0 \) as \( \eta \to 0 \).
2.4 Costs of transitions

The identity of stochastically stable states depends on the transition probabilities of the process. To measure the limiting relative magnitude of these probabilities, a cost function is defined as follows.

**Definition 2.3** The 1-step cost of the process moving from g to g' is defined as:

\[
c(g, g') := \lim_{\eta \to 0} -\eta \log P_\eta(g, g'),
\]

adopting the convention that \(-\log 0 = \infty\).

\(c(g, g')\) is the exponential decay rate of the transition probability from \(g\) to \(g'\). The rarer a transition, the higher its cost. Impossible transitions have infinite cost. Note that for \(g \notin \mathcal{C}\), there is a zero cost transition from \(g\). This is because there is some \(g' \neq g\), such that \(P_\eta(g, g')\) does not approach zero as \(\eta \to 0\). We are also interested in the overall cost of moving between \(g\) and \(g'\), even if many steps are required. Let the \(t\)-step transition probabilities be given by \(P^{t}_\eta(g, g') \equiv P(g^t = g' | g^0 = g, P_\eta(., .))\).

**Definition 2.4** The overall cost of the process moving from \(g\) to \(g'\) is defined as:

\[
C(g, g') := \min_{t \in \mathbb{N}} \lim_{\eta \to 0} -\eta \log P^{t}_\eta(g, g'),
\]

We make one further assumption: we rule out other-regarding mistake probabilities. That is, the cost of a mistake by a pair \((i, j)\) is independent of the current matching of every player other than \(i\) and \(j\). Given that the unperturbed dynamic is self-regarding, this seems a reasonable restriction.

**Assumption 3 (self-regarding mistake probabilities)**

If \(g(i) = g'(i)\) and \(g(j) = g'(j)\), then \(c(g, g - ij) = c(g', g' - ij)\) and \(c(g, g + ij) = c(g', g' + ij)\).

A spanning tree rooted at \(g^* \in \mathcal{C}\) is a directed graph over the set \(\mathcal{C}\) such that every \(g \in \mathcal{C}\) other than \(g^*\) has exactly one exiting edge, and the graph has no cycles. The cost of a spanning tree is the sum of the costs of its edges given by \(C(., .)\). A minimum cost spanning tree is a spanning tree whose cost is lower than or equal to the cost of any other spanning tree. A state \(g^* \in \mathcal{C}\) is stochastically stable only if there exists a minimum cost spanning tree rooted at \(g^*\) (Young, 1993). Finding minimum cost spanning trees can be difficult. The principal contribution of the current paper is to show that any minimum

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7 For many dynamics, ‘only if’ can be replaced by ‘if and only if’. See Sandholm (2010) for details.
8 The same applies to radius-(modified)coradius methods (Ellison, 2000).
cost spanning tree, and hence any stochastically stable matching, must be in the set of matchings which are most robust to one-shot deviation. We call a transition \( g \to g' \) from a matching \( g \in G \) the least cost deviation from \( g \) if it has the lowest cost of all possible 1-step transitions from \( g \).

**Definition 2.5** Denote the set of possible least cost deviations from \( g \in G \) by:

\[
L(g) := \arg\min_{g' \neq g} c(g, g')
\]

and the set of players involved in least cost deviations from \( g \in G \) by:

\[
N_L(g) := \{(i, j) \in M \times W : \exists g' \in L(g) : g' = g - ij \text{ or } g' = g + ij\}
\]

\( c_L(g) \) will be used to denote the cost of the least cost deviation from \( g \).

We use the word deviation as we shall be interested in the application of these concepts to \( g \in C \).

### 3 Stochastically stable matchings

Define \( OS \), the set of matchings which are most robust to one-shot deviation:

\[
OS = \left\{ g \in G : c_L(g) = \max_{g' \in G} c_L(g') \right\}
\]

As \( c_L(g) \) is strictly positive only for \( g \in C \), it must be that \( OS \subseteq C \). We will show that \( OS \) contains \( SS \): a stochastically stable matching must be comparatively robust against one-shot deviation. If \( OS \) contains only one matching, then that matching must be uniquely stochastically stable.

Klaus et al. (2010) show that a single mistake suffices to move from any \( g \in C \) to some other \( g' \in C \). We show that the least cost deviation from a stable matching \( g \not\in OS \) is enough to escape from its basin of attraction, and that the unperturbed dynamic can subsequently lead the process closer to \( OS \subseteq C \). This result is proved in Lemma 3.4, from

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9This differs from the concept of the radius of a stable state \( g \in C \) (Ellison, 2000, citing a no longer extant working paper of Evans, 1993). The radius is defined as \( R(g) = \min_{g' \in C \setminus \{g\}} C(g, g') \) and requires a different stable state to be reached by the process. It turns out that in the problems considered in the current paper \( c_L(g) = R(g) \) for all stable matchings outside of a specific set, but this does not follow from the definitions. 
which the main theorem is proven using a minimal cost spanning tree argument. First, we present a couple of lemmas which assist in the proof of Lemma 3.4.

The following lemma shows that the least cost deviation from a stable matching \( g \notin \text{OS} \) cannot involve two single players forming a partnership.

**Lemma 3.1** Suppose that \( g \in \mathcal{C} \) and \( g \notin \text{OS} \). If \((i,j) \in N_L(g)\), then \( g(i) \neq \emptyset \) and/or \( g(j) \neq \emptyset \).

**Proof.** Suppose \( g(i) = \emptyset \) and \( g(j) = \emptyset \). Then, \( i \) and \( j \) are single in every stable matching (Theorem 2.22 of Roth and Sotomayor (1992)), including the matchings in \( \text{OS} \). As \((i,j) \in N_L(g), g + ij \in L(g)\). Then for \( g^* \in \text{OS}, c_L(g^*) \leq c(g, g + ij) = c(g, g + ij) = c_L(g)\), therefore \( g \in \text{OS} \), which contradicts our premise. \( \blacksquare \)

The next lemma demonstrates that if a pair is involved in a least cost deviation from a stable matching \( g \notin \text{OS} \), then they do not both have the same current partner as in some matching within \( \text{OS} \).

**Lemma 3.2** Suppose that \( g \in \mathcal{C} \) and \( g \notin \text{OS} \). If \((i,j) \in N_L(g)\), then for all \( g^* \in \text{OS}, g(i) \neq g^*(i) \) and/or \( g(j) \neq g^*(j) \).

**Proof.** Let \( g^* \in \text{OS} \). Suppose \( g(i) = g^*(i) \) and \( g(j) = g^*(j) \). If \( g(i) = j \), then \( c_L(g^*) \leq c(g^*, g^* - ij) = c(g, g - ij) = c_L(g) \). If \( g(i) \neq j \), then \( c_L(g^*) \leq c(g^*, g^* + ij) = c(g, g + ij) = c_L(g) \). Therefore \( g \in \text{OS} \), which contradicts our premise. \( \blacksquare \)

We now present the key lemma, which asserts that following the least cost deviation from any stable matching \( g \notin \text{OS} \), the unperturbed dynamic can move to another stable matching which is strictly closer to \( \text{OS} \) than the initial matching. First, we define an index \( m \) which measures the similarity between matchings.

**Definition 3.3** \( m(g, g') \) is the number of players who have the same partner in \( g \) and \( g' \).

\[
m(g, g') := |\{i \in N : g(i) = g'(i)\}|
\]

**Lemma 3.4 (Getting Closer Lemma)** Let \( g^* \in \text{OS} \). Suppose that \( g \in \mathcal{C} \) and \( g \notin \text{OS} \). Let \( g_1 \in L(g) \). Then, \( \exists g' \in \mathcal{C}, t \in \mathbb{N}_+, \text{ such that } m(g^*, g') > m(g^*, g) \) and \( P^t_0(g_1, g') > 0 \).

The proof of Lemma 3.4 is given in the appendix. Using Figure 1 to illustrate our argument, we here emphasize how our result is stronger than existing results in the literature, such as Lemma 5 of Klaus et al., 2010 and a similar claim in Diamantoudi et al. (2004). These results show that, starting from any given unstable matching, it is possible, under the unperturbed dynamic, to reach a stable matching \( g' \in \mathcal{C} \) which is strictly closer to a target stable matching \( g^* \) than the initial unstable matching is to \( g^* \). In Figure 1, this result
Figure 1: $g_1, g^*$ are stable matchings, $g_1, g_2$ are unstable matchings. Distance on the page represents distance under $m(.,.)$. Directed edges represent transitions and the numbers above the edges their respective transition costs.

corresponds to the existence of a zero cost path from $g_1$ and $g_2$ to some stable matching which is closer to $g^*$. From $g_1$, it may be the case that any such path reaches $g$. The deviation from $g$ to $g_1$ may not lead to a stable matching closer to $g^*$. Such a result suffices for the subsequent stochastic stability arguments of Klaus et al., 2010, as all mistakes have the same cost in their model, so from $g$ it is possible to ‘choose’ a desirable deviation such as the one to $g_2$ in Figure 1. From $g_2$, $g^*$ can be reached at no further cost. This option is not open to us. It may be that the least cost deviation from $g$ moves away from $g^*$, such as is the case in the deviation from $g$ to $g_1$ in the figure. We prove, that if $g \notin OS$ and $g^* \in OS$, then there exists a path (the dashed line in the figure) from $g_1$ that circumvents $g$, reaching an unstable state such as $g_2$, which is at least as close to $g^*$ as $g$ is to $g^*$. The application of previous results (generalized to the many to one case by Lemma 5.5 of the current paper) completes the argument. The least cost deviation from $g$ suffices to move the process to a stable matching which is closer to $g^* \in OS$. The example in Section 4.2 will further illustrate these arguments.

Note that Lemma 3.4, and with its many-to-one equivalent later in the paper, can be understood as ‘paths to stability’ results which are stronger than the existing results in the literature. They allow us to say more than could previously be said about which stable states can be reached from different starting points. The knowledge of these paths gained from the Lemma is exactly what is required to prove the main theorem.

**Theorem 3.5** $SS \subseteq OS$.

The formal proof is in the appendix. In brief, any stochastically stable matching must be the root of a minimum cost spanning tree. If a tree is rooted at some $g \in C$, $g \notin OS$, then Lemma 3.4 can be used to build another tree rooted at some $g^* \in OS$. Starting at $g$, use Lemma 3.4 to add edges between stable matchings which get progressively closer to $g^*$. We obtain a sequence $(g = g_1, \ldots, g_L = g^*)$ with edges between $g_i$ and $g_{i+1}$ for $i = 1, \ldots, L - 1$. Each of these new edges has the cost of a lowest cost deviation, $C(g_i, g_{i+1}) = \ldots$
Deleting the edge exiting $g^*$, we are left with a tree rooted at $g^*$. As $g \notin OS$, $g^* \in OS$, the cost of the new edge exiting $g$ must be lower than the cost of the deleted edge which exited $g^*$. So the tree rooted at $g^*$ has a lower total cost than the total cost of the tree rooted at $g$. Therefore no tree rooted at $g$ can be a minimum cost spanning tree. That is, $g \notin SS$.

**Remark 3.6** Consider the special case of uniform mistake probabilities, that is when there exists $a \in \mathbb{R}$ such that for all $g, g' \in G$, $c(g, g') > 0$ implies $c(g, g') = a$. It follows immediately from the proof of Theorem 3.5 that $SS = OS$. The result of Jackson and Watts (2002) and Klaus et al. (2010) is recovered.

Finally, we note that the proof of Theorem 3.5 extends to give a bound on convergence times.

**Remark 3.7** It follows immediately from the proof of Theorem 3.5 that the modified-coradius (see Ellison, 2000) of $OS$ equals $\max_{g \in OS} c_L(g)$ and that therefore, starting from any matching, the expected hitting time of $OS$ is $O(e^{\frac{1}{\eta} \max_{g \in OS} c_L(g)})$.

So, $SS \subseteq OS$. This is important, as the set $OS$ is defined solely by reference to local properties of the stable matchings. Stochastically stable matchings must be matchings which are most robust to one-shot deviation. If $OS$ is a singleton, then the unique stochastically stable state can be determined solely by looking at the lowest cost one-shot deviation from stable states: there is no need to resort to minimal cost spanning trees or to radius-coradius methods.

## 4 Examples

In this section we apply the one-shot deviation principle to study stochastic stability under commonly used choice rules. In Section 4.1, we consider a dynamic induced by the logit choice rule and link $OS$ to the notion of the *least-core* proposed by Maschler et al. (1979). Thus, our one-shot deviation principle combined with logit choice provides an evolutionary foundation for the least-core.¹⁰ In Section 4.2, we provide a comparison of dynamics induced by three leading choice rules, the uniform mistake, the logit choice, and the probit choice.

¹⁰We thank Bary Pradelski and Heinrich Nax for bringing this connection to our attention. Nax and Pradelski (2013) adapt the methods of the current paper to give least-core selection in assignment problems.
4.1 The logit choice rule

At the beginning of period $t + 1$, a pair of players $(i, j)$ is selected at random according to a distribution $F_{g^t}(.)$ with full support on $\mu_g$. $g^{t+1}$ is determined as follows:

(i) If $g^t(i) = j$, then $g^{t+1} = g^t - ij$ with probability

$$1 - \prod_{k \in \{i, j\}} \frac{e^{\frac{1}{\eta}u_k(g')}}{e^{\frac{1}{\eta}u_k(g')} + e^{\frac{1}{\eta}u_k(g' - ij)}}.$$

That is, each of $i$ and $j$ chooses to cut or retain the link $ij$ with probabilities given by the logit choice rule, and unless both players choose to retain the link, it will be cut.

(ii) If $g^t(i) \neq j$, then $g^{t+1} = g^t + ij$ with probability

$$\prod_{k \in \{i, j\}} \frac{e^{\frac{1}{\eta}u_k(g' + ij)}}{e^{\frac{1}{\eta}u_k(g')} + e^{\frac{1}{\eta}u_k(g' + ij)}}.$$

That is, $i$ and $j$ each agree to leave their existing partner and form a new link $ij$ with probability given by the logit choice rule. Both $i$ and $j$ must agree for a new partnership to be formed.

(iii) $g^{t+1} = g^t$ otherwise.

Under the logit choice rule, transition probabilities are sensitive to the amount by which cardinal utility is reduced. The sum of negative changes in revising players’ payoffs for transition $g \rightarrow g'$ is the cost of $g \rightarrow g'$ (Sawa, 2013). If the easiest transition at matching $g$ is for two players to form a partnership, then:

$$c_L(g) = \min_{ij \notin g} \left[ \max\{u_i(g) - u_i(g + ij), 0\} + \max\{u_j(g) - u_j(g + ij), 0\} \right],$$

whereas if the easiest transition at matching $g$ is for a player to dissolve an existing partnership, then:

$$c_L(g) = \min_{i: g(i) \neq \emptyset} \left[ \max\{u_i(g) - u_i(g - ig(i)), 0\} \right] = \min_{i: g(i) \neq \emptyset} \left[ \max\{u_i(g) - u_i(\emptyset), 0\} \right].$$

For the logit choice rule, $c_L(g)$ is therefore the minimum of the quantities in (3) and (4).

Example 4.1 Suppose that $M = \{m_1, m_2, m_3\}$, $W = \{w_1, w_2, w_3\}$, and that the matrix giving their payoffs from a given match is shown below. For example, the top left cell tells us that $m_1$ gets
a payoff of 10 from being matched with \( w_1 \). The payoffs from being single are zero for all men and women. Let the perturbed dynamic be the logit choice rule.

<table>
<thead>
<tr>
<th></th>
<th>( w_1 )</th>
<th>( w_2 )</th>
<th>( w_3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( m_1 )</td>
<td>10,1</td>
<td>5,5</td>
<td>1,10</td>
</tr>
<tr>
<td>( m_2 )</td>
<td>1,10</td>
<td>10,1</td>
<td>5,5</td>
</tr>
<tr>
<td>( m_3 )</td>
<td>5,5</td>
<td>1,10</td>
<td>10,1</td>
</tr>
</tbody>
</table>

There are three stable matchings as below.

\[
g_1 = \{m_1w_1, m_2w_2, m_3w_3\}, \quad g_2 = \{m_1w_2, m_2w_3, m_3w_1\}, \quad g_3 = \{m_1w_3, m_2w_1, m_3w_2\}.
\]

Note that \( g_1 \) is man-optimal and \( g_3 \) is woman-optimal. Also note that \( c_L(g) = 1 \) for \( g \in \{g_1, g_3\} \). For example, one of the least cost deviations from \( g_1 \) is \( w_1 \) becoming single, which costs 1. Let \( g' \) denote the resulting matching. The cost of this deviation is:

\[
c_L(g_1) = c(g_1, g') = u_{w_1}(g_1) - u_{w_1}(g') = 1 - 0 = 1.
\]

Followed by \( \{m_1w_2\}, \{m_2w_3\}, \{m_3w_1\} \) matching sequentially, the dynamic will reach \( g_2 \).

Moreover, \( c_L(g_2) = 4 \). One of the least cost deviations from \( g_2 \) is for \( m_1 \) and \( w_1 \) to partner, causing the payoff of \( w_1 \) to decrease by 4. These values for \( c_L(.) \) imply that \( OS = \{g_2\} \). So \( SS = \{g_2\} \), the unique stochastically stable matching is \( g_2 \).

Under the logit choice rule, \( OS \) corresponds to a non-transferable utility version of the least-core (Maschler et al., 1979).

**Definition 4.2** For \( A \subseteq N \), let \( G(A, g) \) be the set of matchings \( g' \neq g \) such that \( ij \in g' \) for all \( i, j \notin A \), \( ij \in g \), and \( ij \notin g' \) for all \( i \notin A, ij \notin g \). Then the excess of \( A \) at \( g \) is defined as

\[
e(A, g) := \max_{g' \in G(A, g)} \sum_{i \in A} \min\{0, u_i(g') - u_i(g)\},
\]

and the least-core is

\[
\mathcal{LC} := \arg\min_{g \in \mathcal{C}} \max_{A \neq \emptyset} e(A, g).
\]

Note that in contrast to the definition of excess in Maschler et al. (1979), we do not allow players’ gains to enter the calculation. Within the core, excess is a measure of the amount

\footnotetext{11}{This example is a version of Example 2.17 of Roth and Sotomayor (1992) in which we have removed a man and a woman.}
by which a constraint is satisfied, and in a non-transferable utility setup this is unaffected
by potential gains in payoff. In marriage problems, the maximum excess can be found by
analyzing \( A \) such that \( |A| \leq 2 \). The following proposition follows immediately.

**Proposition 4.3** Under the logit choice rule, \( OS = \mathcal{L}C \).

We can further characterize properties of matchings in \( OS \) for generic payoffs un-
der logit and similar choice rules. Suppose a set of payoff vectors which satisfy, for all
\( i,j,i',j' \in N, g, g' \in G \),

\[
u_i(g \pm ij) - u_i(g) = u_{i'}(g' \pm i'j') - u_{i'}(g') \Rightarrow i = i', j = j', g(i) = g'(i').
\]

where \( g \pm xy \) is \( g + xy \) if \( xy \notin g \) and \( g - xy \) otherwise. The payoffs are generic in the sense
that the complement of the closure of such a set has Lebesgue measure zero in \( \mathbb{R}^{|G| \times |N|} \)
satisfying payoff assumptions in Section 2.1.

**Remark 4.4** For generic payoffs, under the logit choice rule, the set of pairs of players involved
in least cost deviations is a singleton, and is identical across matchings in \( OS \). That is, \( N_L(g) = N_L(g') \).

\( g, g' \in OS \). This implies that \( g(i) = g'(i) \) and \( g(j) = g'(j) \) for \( (i, j) \in N_L(g) \) for
all \( g, g' \in OS \).

### 4.2 Comparison of alternative rules

By means of an example, we now consider three popular choice rules successively. In doing so we show differences and subtleties, highlighting the difference between the
current work and previous work which considers only uniform errors (Jackson and Watts,
2002; Klaus et al., 2010).

Suppose that \( M = \{m_1, m_2, m_3\} \), \( W = \{w_1, w_2, w_3\} \), and that the matrix giving play-
ers’ payoffs from a given match is given below. The stable matchings are \( g_W = \{m_1w_2, m_2w_1, m_3w_3\} \) and \( g_M = \{m_1w_1, m_2w_2, m_3w_3\} \). \( g_W \) is the woman optimal matching
and \( g_M \) the man optimal matching.

\[
\begin{array}{ccc}
  & w_1 & w_2 & w_3 \\
 m_1 & 10,5 & 5,10 & 6,1 \\
m_2 & 5,10 & 10,5 & 2,10 \\
m_3 & 3,3 & 2,3 & 5,5 \\
\end{array}
\]
4.2.1 Uniform mistakes

Consider a move from $g_W$ to $g_M$. Earlier work on uniform mistakes would consider the simplest sequence of transitions from $g_W$ to $g_M$. For example, Step 2 in the proof of Theorem 2 of Klaus et al. (2010) could be interpreted as follows. Since $w_2 = g_W(m_1) \neq g_M(m_1) = w_1$, the pair $(m_1, w_1)$ is chosen to match. This is a mistake as $w_1$ loses payoff. The resulting matching is $g_2 = \{m_1w_1, m_2w_2, m_3w_3\}$. Note that $m(g_2, g_M) > m(g_W, g_M)$. That is, the new matching is closer to $g_M$ than was the original matching. From $g_2$, the unperturbed dynamic can costlessly reach $g_M$. One mistake which is carefully chosen to increase $m(\cdot, g_M)$ can move the process from $g_W$ to $g_M$. A similar trick is used to move from $g_M$ to $g_W$, and as every mistake has the same cost under uniform errors, $OS = \{g_W, g_M\}$. Further, as discussed in Remark 3.6, $SS = \{g_W, g_M\}$. Note that when both players are mistaken in forming a match, it does not matter whether this counts as one or two errors: as we have just seen, on the required paths, it is never the case that both players who form a match are mistaken in doing so.

4.2.2 Logit choice rule

When employing payoff-dependent (or otherwise differing) mistakes, we are treading into more tricky territory. Theorem 3.5 shows we can, indeed must, restrict our attention to particular deviations which incur the minimum cost. The least cost deviation from $g_W$ under the logit choice rule has $\{m_2w_3\}$ forming a link. Denote the resulting matching $g_1 = \{m_1w_2, m_2w_3, w_1, m_3\}$. $m_2$ loses 3 units of payoff from the match, and $w_3$ does not make a loss, so $c(g_W, g_1) = 3 + 0 = 3$. Observe that $g_1$ is more distant from $g_M$ than was the original matching: $m(g_1, g_M) < m(g_W, g_M)$. What Lemma 3.4 shows is that even from $g_1$, the process can costlessly reach some state closer to $g_M$ than $g_W$ is. Figure 1 earlier in the paper illustrates the contrast between the paths used when considering uniform errors and the paths which must be used to prove Theorem 3.5. In the example under consideration, following the move to $g_1$, it can be the case that $\{m_1w_1\}, \{m_2w_2\}, \{m_3w_3\}$ match sequentially, reaching $g_M$. These transitions have zero cost. So, a single least cost deviation suffices to move the process from $g_W$ to $g_M$. While, the least cost deviation from $g_M$ involves $m_3$ and $w_1$ matching. They each make a payoff loss of 2 by forming this match, so the cost of the mistake is $2 + 2 = 4$. We conclude that $OS = \{g_M\}$, so $g_M$ is the unique stochastically stable matching.
4.2.3 Probit choice rule

The reader may conjecture that because of our one-shot deviation principle, any mistake model in which the cost of mistakes is increasing in payoff loss will always give the same result as the logit choice rule. Such a conjecture is false, as we now show. The conjecture is true if we choose an error model in which every mistake only involves loss of payoff by a single player. However, the possibility of two players erring at the same time complicates matters. To see this, consider the probit choice rule. Specifically, consider a perturbed dynamic similar to that in Section 4.1, differing in that players decide whether to accept new matchings or leave existing matchings according to probit instead of logit choice. Dokumaci and Sandholm (2011) show that under probit choice, mistake costs are proportional to the square of payoff loss. However, the combined cost of two players making a mistake will still be additive. The example of this section has been constructed so that least cost deviations remain the same under logit and probit. Calculating, we see that $c_{L}(g_W) = 3^2 + 0^2 = 9$ and $c_{L}(g_M) = 2^2 + 2^2 = 8$. That is, under probit choice, OS = \{g_W\}, so $g_W$ is the unique stochastically stable matching. Note that compared to logit, the convexity of costs in probit is friendly towards mistakes by multiple players.\[12\]

More generally, OS under the probit choice rule coincides with a variant $\mathcal{LC}_{pro}$ of the least-core (Definition 4.2) with excess renormed as

$$e_{pro}(A,g) := \max_{g' \in G(A,g)} - \left( \sum_{i \in A} \left[ \min\{0, u_i(g') - u_i(g)\} \right] ^2 \right)^{\frac{1}{2}}.$$ 

Recalling that $-e(A,g)$ is a minimum sum of potential payoff losses of members of $A$, $-e_{pro}(A,g)$ can be interpreted as the minimum Euclidean length of vectors of potential payoff losses. Hence the difference between logit and probit when it comes to determining OS is equivalent to the difference between using the taxicab\[13\] and Euclidean norms to assess the size of a vector of payoff losses.

4.2.4 Non-genericity of predictions of uniform mistake models

Consider a choice rule with an arbitrary $a \geq 0$ for which individual mistake costs are proportional to payoff loss to the power of $a$. In the special cases of $a = 0, 1, 2$, we have the uniform, logit and probit cases respectively. Considering our example above, if $b = \log 2 / \log 2 - \log 3$, then OS = \{g_M\} for $0 < a < b$, and OS = \{g_W\} for $a > b$. If $a = 0$ or $a = b$,

\[12\]So when it comes to making mistakes, the adage ‘misery loves company’ is more true under probit choice than under logit.

\[13\]The taxicab norm sums the elements of a vector.
then $OS = \{g_W, g_M\}$. Results for the uniform case are non-generic in this class of models.

4.2.5 $OS \neq SS$

Least cost deviations are always enough to move from stable matchings outside $OS$ towards stable matchings in $OS$, but the same is not necessarily true when moving between matchings in $OS$. This fact implies that there exist cases in which $OS \neq SS$.

Let $w_3$ be a slightly eccentric individual who is particularly prone to leaving her partner.\textsuperscript{14} Let the cost of any mistake in which $w_3$ leaves her partner be 1. Let the cost of all other mistakes be 2. Then a least cost deviation from both $g_W$ and $g_M$ is for $w_3$ to leave $m_3$ and become single at a cost of 1. Therefore $OS = \{g_W, g_M\}$. Starting from $g_W$, following this least cost deviation, $\{m_1w_3\}, \{m_2w_2\}, \{m_1w_1\}, \{m_3w_3\}$ can match sequentially at zero cost, reaching $g_M$. However, starting from $g_M$, following the least cost deviation, the only costless transition is for $\{m_3, w_3\}$ to rematch, returning to $g_M$. To move from $g_M$ to $g_W$ a more costly error is required. Hence $SS = \{g_M\} \neq OS$.

5 Many-to-one matching problems

We extend our analysis to many-to-one matching problems, also known as college admissions problems. The difference from one-to-one matching problems is that each agent of one population, the colleges, may be matched with more than one agent of the other population, the students. Each student is matched with at most one college.

There are two sets, $K = \{K_1, \ldots, K_l\}$ and $S = \{s_1, \ldots, s_m\}$, of colleges and students respectively. There is positive integer $q_K$, called the quota, of college $K$ which indicates the maximum number of positions college $K$ may fill. That is, $|g(K)| \leq q_K$. All $q_K$ positions of college $K$ are identical. The set of matchings in the college admissions problem is:

$$G = \{g \in G : (\forall ij \in g, i \in S \Leftrightarrow j \in K), (\forall i \in S, |g(i)| \leq 1), (\forall K_j \in K, |g(K_j)| \leq q_{K_j})\}.$$  

The preferences of college $K$ are determined by the subset of students to which $K$ is matched. That is, although $g(K)$ can now be of size greater than one, it is still the case that $g(K) = g'(K)$ implies that $u_K(g) = u_K(g')$. Preferences over subsets of students are still assumed to be strict: $g(K) \neq g'(K) \Leftrightarrow u_K(g) \neq u_K(g')$.

**Definition 5.1** A matching $g$ is in the core of a matching problem, denoted $g \in C$ if $\nexists A \subseteq N$, $g' \in G$ such that:

\textsuperscript{14}We assume such an individual so as to use the same example and make the discussion simpler. Examples for which $OS \neq SS$ can be shown for logit and probit dynamics.
(i) \(i \notin A, j \notin A, ij \in g \Rightarrow ij \in g'\)

(ii) \(ij \notin g, ij \in g' \Rightarrow i \in A, j \in A\)

(iii) \(i \in A \Rightarrow u_i(g') > u_i(g)\).

We restrict our attention to responsive preferences (Roth, 1985). If a college has responsive preferences, then its preferences over any two students \(s_i, s_j\) are independent of the other students to whom it is matched. That is, if a college, \(K\), prefers \(s_i\) to \(s_j\), and \(T, |T| < q_K\), is some subset of students which includes neither \(s_i\) nor \(s_j\), then the college prefers \(T \cup s_i\) to \(T \cup s_j\). We assume that all colleges have responsive preferences.

**Definition 5.2** The preferences of college \(K \in K\) over sets of students are responsive if they satisfy the following conditions.

(I) If \(g(K) = g'(K) \cup \{s_i\} \setminus \{s_j\}, s_i \notin g'(K), s_j \in g'(K),\) then \(u_K(\{Ks_i\}) > u_K(\{Ks_j\}) \Leftrightarrow u_K(g) > u_K(g')\).

(II) If \(g(K) = g'(K) \cup \{s_i\}, s_i \notin g'(K),\) then \(u_K(\{Ks_i\}) > u_K(\emptyset) \Leftrightarrow u_K(g) > u_K(g')\).

Following Chapter 5 of Roth and Sotomayor (1992), we consider a related marriage problem, in which each college \(K\) is broken into \(q_K\) positions of itself: \(k_1, \ldots, k_{q_K}\), each of which has a quota of one. In the related market, the players are students and college positions each of which has a quota of one. The college positions are assumed to have the same preferences over the individual students as their original college. With a slight abuse of notation, we let \(K\) denote the set of positions in college \(K\), i.e. \(K = \{k_1, \ldots, k_{q_K}\}\), \(g(K) = \bigcup_{1 \leq i \leq q_K} g(k_i)\).

We assume that both the unperturbed and the perturbed dynamics forbid positions of the same college from competing for students. In the original problem, college \(K\) is indifferent between a student filling position \(k_i \in K\) or \(k_j \in K\). Hence it is unrealistic that two positions within the same college compete with one another for a student. We prevent such competition by imposing the assumption below.

**Assumption 4** Let \(v(g) := \{(i,k_j) : k_j \in K, i \neq g(k_j), i \in g(K)\}\). At the beginning of period \(t+1\), the updating pair of players will be chosen according to a distribution with full support on \(\mu \setminus v(g')\).

The next lemma shows that the set of absorbing states, \(\mathcal{C}\), of the dynamic in section 2.2 amended to satisfy Assumption 4 corresponds to the core, \(\mathcal{C}'\), in the college admission problem.

---

15There may exist cases in which different departments of a college compete for students. In such cases, we let \(K\) and \(K'\) be such that \(K \neq K'\) represent different departments.
Lemma 5.3 For \( g' \in G \), let \( g \in G \) satisfy: for all \( K = \{k_1, \ldots, k_{q_K}\} \in K \),

(a) \( i \in g'(K) \iff \exists k_j \in K : \ g(k_j) = i \).

(b) \( \forall i \in N, |g(i)| \leq 1 \).

Then \( g' \in \mathcal{C}' \iff g \in \mathcal{C} \).

Note that the logit dynamic under Assumption 4 is still irreducible. For any network \( g \), a pair \( (g(k), k) \) will cut their link with positive probability in the perturbed dynamic satisfying Assumption 4. For all \( g, g' \in G \), there is a positive probability that the perturbed dynamic starting with \( g \) will become the empty network within \( |N| \) periods, and then will become \( g' \) within another \( |N| \) periods.

Definition 5.4 Define the set of matchings equivalent to \( g \) as:

\[
\text{Eq}(g) = \{g' \in G : g'(K) = g(K) \ \forall K \in K\}.
\]

In words, \( \text{Eq}(g) \) is the set of matchings in which students are matched to the same colleges as they are in matching \( g \), i.e. matchings in \( \text{Eq}(g) \) are identical in the original college admission problem.

We make a natural symmetry assumption on the dynamic regarding the behavior of positions of a college. We assume that the cost of transitions is unaffected by the labelling of the positions of any given college.

Assumption 5 If \( \tilde{g} \in \text{Eq}(g) \); \( k_1, k_2 \in K_i : g(k_1) = \tilde{g}(k_2) \); \( s \in S : g(s), \tilde{g}(s) \in K_j \in K \) or \( g(s) = \tilde{g}(s) = \emptyset \); then:

(i) \( c(g, g + k_1s) = c(\tilde{g}, \tilde{g} + k_2s) \),

(ii) If \( g(k_1) \neq \emptyset \), then \( c(g, g - k_1g(k_1)) = c(\tilde{g}, \tilde{g} - k_2\tilde{g}(k_2)) \), and

(iii) If \( g(s) \neq \emptyset \), then \( c(g, g - sg(s)) = c(\tilde{g}, \tilde{g} - s\tilde{g}(s)) \).

Note that the logit choice rule satisfies Assumption 5.

Take any unstable matching \( g \notin \mathcal{C} \) and a target stable matching \( g' \in \mathcal{C} \). The following lemma, which is important to the results of the paper, shows that, starting from \( g \), the unperturbed dynamic can move to some matching \( \tilde{g}_T \) which is strictly closer to \( g' \) than \( g \) is. This lemma extends the implications of Lemma 5 of Klaus et al. (2010) to many-to-one matching problems. First, define a similarity function for the many-to-one matching problem:

\[
\bar{m}(g, g') := \max_{\tilde{g} \in \text{Eq}(g')} m(g, \tilde{g}) \tag{5}
\]
Note that \( \bar{m}(g, g') \geq m(g, g') \). Also note that \( m(., .) \equiv \bar{m}(., .) \) for one-to-one matching problems.

**Lemma 5.5** Assume Assumption 4 holds. Let \( g \notin \mathcal{C}, g' \in \mathcal{C} \). Then, \( \exists T \in \mathbb{N}_+, g_T \in G, \) such that \( P^T_0(g, g_T) > 0 \) and \( \bar{m}(g_T, g') > \bar{m}(g, g') \).

The proof is left to the appendix, and makes use of the fact that given \( g \) and \( g' \), there is no student matched to different positions of the same college under \( g \) and the \( g^* \) which solves the maximization in (5).

**Lemma 5.6** Let \( g \notin \mathcal{C}, g' \in \mathcal{C} \). Let

\[
g^* \in \arg \max_{\hat{g} \in \operatorname{Eq}(g')} m(g, \hat{g}).
\]

For all \( i \in \mathcal{S}, g(i) \in \mathcal{K}, g^*(i) \in \mathcal{K} \Rightarrow g(i) = g^*(i) \).

**Proof.** Assume \( i \in \mathcal{S}, g(i) \in \mathcal{K}, g^*(i) \in \mathcal{K} \neq g^*(i) \). Let \( g^{**} = g^* + ig(i) + g^*(i)g^*(g(i)) \). Then \( g^{**} \in \operatorname{Eq}(g^*) = \operatorname{Eq}(g') \) and \( m(g, g^{**}) \geq m(g, g^*) + 2 \), contradicting the definition of \( g^* \). □

The proof of Lemma 5.5 relies on the construction of closed cycles of players who have strict preferences between \( g \) and \( g^* \). Lemma 5.6 ensures that players who have the same partner in \( g \) and \( g^* \), and who are therefore indifferent between the two matchings, form separate cycles of size two.

Lemma 5.5 directly implies the following corollary. It is similar to Roth and Vande Vate (1990) except that we have not assumed students to have strict preferences over the positions within colleges.\textsuperscript{16,17}

**Corollary 5.7 (Random paths to stability)** Suppose a college admission problem, its related marriage problem, and an unperturbed dynamic satisfying Assumption 4. For any \( g \notin \mathcal{C} \), there exists \( T \in \mathbb{N}_+, g^* \in \mathcal{C} \), such that \( P^T_0(g, g^*) > 0 \).

Define \( c_L(g) \) and \( OS \) as in the one-to-one matching problem. Using Lemma 5.5, a many-to-one version of Lemma 3.4 can be proved. Then, we have the following theorem. See Appendix for proofs.

**Theorem 5.8** Under Assumptions 4 and 5, \( SS \subseteq OS \).

\textsuperscript{16}See Chapter 5 of Roth and Sotomayor (1992) for a way to construct strict preferences in such problems.

\textsuperscript{17}As students only ever match with a single college, their preferences are substitutable. Therefore, the many-to-many paths to stability result of Kojima and Unver (2008) also implies this corollary.
Example 5.9 (the logit choice rule and Assumption 4)

The implications of Assumption 4 for the costs of deviations under the logit choice rule are as follows. Let

\[ E(g) = \{ik_j : ik_j \notin g, ik_j \in g' \text{ for some } g' \in Eq(g)\}. \]

In words, \( ik_j \in E(g) \) means that \( i \) and \( k_j \in K \) are not matched in \( g \), but \( i \) is matched with some \( k_l \in K, k_l \neq k_j \). Assumption 4 implies that pairs in \( E(g) \) may not deviate if the process is in \( g \). For the logit dynamic, expressions for \( c_L(g) \) will be as in expressions (3) and (4), but with the minimum being taken over \( ij \notin g \cup E(g) \) in expression (3).

We conclude this section with three examples. Example 5.10 is an application of Theorem 5.8 with a note emphasizing the role of Assumption 4. Examples 5.11 and 5.12 demonstrate that extending our result beyond responsive preferences, for example to settings in which colleges have a preference for some form of diversity in their student bodies, is not straightforward.

Example 5.10 Let \( S = \{s_1, s_2, s_3\}, K = \{K, K'\}, K = \{k_1, k_2\} \text{ and } K' = \{k_3\} \). Assume that a college’s utility is additive over the utility it obtains from each student, and that the perturbed dynamic is the logit choice rule. Preferences are given by the following matrix. The payoff from remaining unmatched is assumed to be zero.

\[
\begin{array}{ccc}
  & k_1 & k_2 \\
 s_1 & 10,10 & 10,10 & 5,10 \\
 s_2 & 10,8 & 10,8 & 6,8 \\
 s_3 & 5,9 & 5,9 & 10,5 \\
\end{array}
\]

Observe that under Assumption 4, the set of stable matchings is \( C = \{g_1, g_2, g_3, g_4\} \) where

\[
\begin{align*}
g_1 &= \{(s_1,k_1), (s_2,k_2), (s_3,k_3)\}, & g_2 &= \{(s_1,k_2), (s_2,k_1), (s_3,k_3)\}, \\
g_3 &= \{(s_1,k_1), (s_2,k_3), (s_3,k_2)\}, & g_4 &= \{(s_1,k_2), (s_2,k_3), (s_3,k_1)\}.
\end{align*}
\]

The first two matchings are equivalent, \( g_2 \in Eq(g_1) \). \( s_1 \) and \( s_2 \) are matched to \( K \) in both matchings. Similarly, \( g_4 \in Eq(g_3) \).

Suppose that the current network is \( g_1 \). In the absence of Assumption 4, a deviation by \( (s_1,k_2) \) to \( g_1 + s_1k_2 \) could occur with cost zero. Subsequently, \( (s_2,k_3) \) and \( (s_3,k_1) \) could form partnerships, and the process could reach \( g_4 \) without any additional cost. So \( C(g_1, g_4) \) would equal zero. Similarly, we can cycle between all of the matchings in \( C \).
Under Assumption 4 \((s_1, k_2)\) will never be selected as a revising pair when the current state is \(g_1\). The least cost deviation from \(g_1\) is \(L(g_1) = \{g_1 + s_2k_3\}\) with cost \(c_L(g_1) = 4\). Also, \(c_L(g_2) = 4\), \(c_L(g_3) = 1\), \(c_L(g_4) = 1\). \(OS = \{g_1, g_2\}\). Since \(g_1\) and \(g_2\) are equivalent, the unique stochastically stable matching is that \(K\) and \(K'\) are matched to \(\{s_1, s_2\}\) and \(s_3\) respectively.

**Example 5.11 (Non-responsive preferences)** Consider a college admissions problem with \(S = \{s_1, s_2, s_3\}\), \(K = \{K\}\) and \(q_K = 3\). Let \(u_x(K) = 10\) and \(u_x(\emptyset) = 0\) for all \(x \in S\). Also, for \(X \subseteq S\), let

\[
u_K(X) = \begin{cases} 
5 & \text{if } |X| = 3, \\
4 & \text{if } |X| = 2, \\
-2 & \text{if } |X| = 1, \\
0 & \text{if } X = \emptyset.
\end{cases}
\]  

(6)

In words, each student prefers being in \(K\) to being out. College \(K\) prefers to have at least two students to none, but prefers none to having one student only. Let the perturbed dynamic be the logit choice rule.

There are two stable matchings: one where \(K\) accepts all students and another where none are accepted. The former is uniquely stochastically stable, while the latter is most robust to one-shot deviation. To see this, observe that one costly deviation, which costs 2, is enough to move from none accepted to all accepted. While at least two costly deviations, which cost 7, are required in the opposite move. Lemma 5.5 does not apply here and hence Theorem 5.8 does not hold.

Note that in the absence of responsive preferences, Lemma 5.3 does not hold. In Example 5.11, we just saw that although the core contains a unique matching in which every student matches with the college, there is an additional stable matching in which no students match with the college. It may be argued that, in the absence of the equivalence result of Lemma 5.3, a richer process of strategic updating should be used. The following example allows groups of players to rematch amongst themselves each period, with no limitations on the size of such a group. One of the colleges has a preference for a diverse student body. This violates responsiveness, and Theorem 5.8 still fails to hold.

**Example 5.12 (Calibrate diversity)** Consider a college admissions problem with three students \(S = \{s_{y1}, s_{y2}, s_{p3}\}\), and two colleges \(K = \{K_y, K_p\}\), \(q_{K_y} = q_{K_p} = 3\). The students are either Yellow students \((s_{y1}, s_{y2})\), or Pink students \((s_{p3})\). Yellow students prefer college \(K_y\), and Pink students prefer college \(K_p\). Let \(u_{s_{y1}}(K_y) = u_{s_{y2}}(K_y) = u_{s_{p3}}(K_p) = 20\), \(u_{s_{y1}}(K_p) = u_{s_{y2}}(K_p) = u_{s_{p3}}(K_y) = 10\), \(u_x(\emptyset) = 0\) for all \(x \in S\). Let the utilities of the colleges, which do not satisfy responsiveness, be given by the following table.
\[
\begin{array}{c|cc}
X \subseteq S & u_{K_y}(X) & u_{K_p}(X) \\
\emptyset & 0 & 0 \\
\{s_{y1}\}, \{s_{y2}\}, \{s_{p3}\} & -1 & -2 \\
\{s_{y1}, s_{p3}\}, \{s_{y2}, s_{p3}\} & 2 & 2 \\
\{s_{y1}, s_{y2}\} & -5 & 2 \\
\{s_{y1}, s_{y2}, s_{p3}\} & 5 & 5 \\
\end{array}
\]

Both colleges would like to have all of the students and prefer no students to having one student only. College \(K_p\) prefers two students to none. College \(K_y\) prefers two students to none only if the students are of different types. That is, college \(K_y\) has a strong antipathy towards a non-diverse student population. The core of the problem contains two matchings, \(g_y\) and \(g_p\), in which all of the students attend either college \(K_y\) or college \(K_p\) respectively. That is,

\[
g_y(K_y) = \{s_{y1}, s_{y2}, s_{p3}\} = g_p(K_p).
\]

Let the process of rematching be as follows. Each period some subset of players \(A \subseteq N\) is chosen. Let \(g^t = g\) be the current matching. A conjectured rematching \(g'\) which satisfies (i) and (ii) of Definition 5.1 is chosen at random and accepted by each member of \(A\) with probabilities given by the logit choice rule. If any member of \(A\) rejects the rematching, then \(g^{t+1} = g^t = g\). If every member of \(A\) accepts the rematching, then \(g^{t+1} = g'\).

From \(g_p\), the least cost deviation involves student \(s_{y1}\) or \(s_{y2}\) moving to college \(K_y\). That is, \(L(g_p) = \{g_1, g_2\}\), \(g_1(K_y) = \{s_{y1}\}\), \(g_1(K_p) = \{s_{y2}, s_{p3}\}\), \(g_2(K_y) = \{s_{y2}\}\), \(g_2(K_p) = \{s_{y1}, s_{p3}\}\). Note that \(c_L(g_p) = 1\). However, following such a deviation, there is no zero cost path to \(g_y\), and unless there is at least one further, considerably more costly, deviation (such as college \(K_y\) accepting another \(y\)-student), \(g_y\) cannot be reached.

Observe that from \(g_y\), the least cost deviation involves student \(s_{p3}\) and college \(K_p\) matching. That is, \(L(g_y) = \{g_3\}\), \(g_3(K_y) = \{s_{y1}, s_{y2}\}\), \(g_3(K_p) = \{s_{p3}\}\), and \(c_L(g_y) = 2\). Following such a deviation, a zero cost path to \(g_p\) exists.

So we have that \(OS = \{g_y\}\), yet \(SS = \{g_p\}\). Theorem 5.8 does not hold.

### 6 Roommate problems

In the one-sided matching problem, or roommate problem, the set of admissible matchings is not restricted to be bipartite. Anyone can partner with anyone. The set of networks of interest is broadened to:

\[
G = \{g \in \mathcal{G} : (\forall i \in N, |g(i)| \leq 1)\}.
\]
Gusfield and Irving (1989) show that two key properties of marriage problems extend to all roommate problems with strict preferences over partners. Firstly, the set of unmatched players is the same at every stable matching. Secondly, if \( g, g' \in C \), \( i \) prefers \( g \) to \( g' \), \( g(i) = j \), \( g'(i) = k \neq j \), then both \( j \) and \( k \) prefer \( g' \) to \( g \). These properties are exactly those used in our results of section 3. Furthermore, Diamantoudi et al. (2004) show that if \( C \) is nonempty, then there exists a sequence of mutually beneficial blockings ending in \( C \). In the context of this paper, this means that nonempty \( C \) implies that all recurrent classes of the unperturbed Markov process lie in \( C \). There are no absorbing cycles. Assuming nonemptiness of \( C \), lemmas 3.1, 3.2, 3.4 still hold. It follows that:

**Theorem 6.1** If \( C \neq \emptyset \), then \( SS \subseteq OS \).

Thus our main result does not rely on two-sidedness of the matching market.

7 Conclusion

This paper has shown that in marriage problems, roommate problems and college admission problems, all stochastically stable matchings are in the class of matchings which are most robust to one-shot deviation. There are two significant implications of this from a market design perspective. Firstly, a desired matching may not be stochastically stable, so even if implemented in the short run, in a world in which people make the occasional mistake, it would be rarely observed in the long run. Secondly, making a desired matching more robust to one-shot deviation than any other matching will suffice to make it uniquely stochastically stable. The main results, which link stochastic stability to a local property of the individual matchings, are derived from the structure of stable matchings and from the unperturbed blocking dynamic. The class of unperturbed blocking dynamics we use is common in the paths to stability literature. Further attempts to extend our results to, for example, hedonic games or many-to-one matchings with complementarities, are left for future work.

A Appendix

In this section, we prove Theorem 5.8. Theorem 3.5 is implied by Theorem 5.8, as Assumptions 4 and 5 do not have any effect in the one-to-one setting. Similarly, Lemma 3.4 is implied by its many-to-one equivalent, Lemma A.4. The proof of Lemma 5.5, one of our key lemmas, is given as below.
Proof of Lemma 5.5. Let
\[ g^* \in \arg\max_{\hat{g} \in \mathcal{E}_q(g')} m(g, \hat{g}). \]

If there exists \( i \in N \) such that \( g(i) \neq \emptyset \) and \( c(g, g - ig(i)) = 0 \) and \( g^*(i) = \emptyset \), then let \( g_T = g - ig(i) \) and we are done: \( \bar{m}(g_T, g') \geq m(g_T, g^*) > m(g, g^*) = \bar{m}(g, g') \).

If there does not exist such an \( i \in N \), let each \( i \in N \) such that \( u_i(\emptyset) > u_i(g) \) leave their partners. Denote the resulting matching \( g_1 \). Note that \( m(g_1, g^*) = m(g, g^*) \). \( g_1 \notin \mathcal{C} \) as if \( g_1 \neq g \), for \( i \in S \) such that \( g(i) \neq g_1(i) = \emptyset \), \( g^*(i) \neq \emptyset \), so \( i \) is not single in any stable matching. Note that \( g^* \in \arg\max_{\hat{g} \in \mathcal{E}_q(g')} m(g_1, \hat{g}) \). As \( g_1 \notin \mathcal{C} \), \( \exists (i, k_j) : c(g_1, g_1 + ik_j) = 0 \).

Case I: \( \exists (i, k_j) : c(g_1, g_1 + ik_j) = 0 \) and \( ik_j \in g^* \).

Let \( g_T = g_1 + ik_j \). Then \( \bar{m}(g_T, g') \geq m(g_T, g^*) > m(g_1, g^*) = m(g, g^*) = \bar{m}(g, g') \) and we are done.

Case II: \( \forall (i, k_j) : c(g_1, g_1 + ik_j) = 0 \), \( ik_j \notin g^* \).

First, we decompose the player set \( N \) into singletons who are unmatched in \( g_1 \) and \( g^* \), pairs of players who have the same partner in \( g_1 \) and \( g^* \), and cycles defined below. Then, we will construct a path of blockings which increase \( \bar{m}(\cdot, g^*) \).

For all \( i \in S \): \( g_1(i) \in K, g^*(i) \in K^*, K = K^* \), we have by Lemma 5.6 that \( g_1(i) = g^*(i) \).

For all \( i \in S \): \( g_1(i) \in K, g^*(i) \in K^*, K \neq K^* \), either \( u_i(g_1) > u_i(g^*) \) or \( u_i(g^*) > u_i(g_1) \). We assume that \( u_i(g_1) > u_i(g^*) \). The arguments when the converse holds are identical.

Let \( f : N \to N \) be such that \( f(j) = g_1(j) \) if \( u_i(g_1) > u_i(g^*) \) and \( f(j) = g^*(j) \) otherwise.\(^{18}\)

Suppose a sequence \( \{i, f(i), f^2(i), f^3(i), \ldots\} \) where \( f^2(i) = f(f(i)) \) and \( f^k(\cdot) \) for \( k \geq 3 \) is defined similarly. Since \( N \) is finite, the sequence must repeat and create a cycle. Denote the cycle by a sequence \( (n_1, n_2, \ldots, n_m) \), where \( n_1 = i, n_1 = f^{l-1}(n_1) \), and \( n_m \) is the last non-repeated element of the cycle. In the sequence, members’ preferences alternate between \( g_1 \) and \( g^* \), i.e. \( g^*(n_j) = n_{j+1} \) if \( j \) is odd, and \( g_1(n_j) = n_{j+1} \) otherwise.\(^{19}\)

Note that \( m \) is even and that \( g^*(n_m) = i \) under the assumption that \( u_i(g_1) > u_i(g^*) \). Thus, \( N \) can be decomposed into singletons, pairs of players and cycles in which players have different partners in \( g_1 \) and \( g^* \).

Now, observe that \( \#(i, k_j) : c(g_1, g_2 = g_1 + ik_j) = 0 \), \( u_i(g_1) \geq u_i(g^*) \) and \( u_{k_j}(g_1) \geq \)

\(^{18}\)Note that strict preferences and the definition of \( g^* \) imply that \( u_i(g) \neq u_i(g^*) \) if \( g(j) \neq g^*(j) \).

\(^{19}\)If \( g_1(n_j) = n_{j+1} \), then \( n_j \) prefers \( g_1 \) to \( g^* \), so \( n_{j+1} \) cannot prefer \( g_1 \) to \( g^* \), or \( (n_j, n_{j+1}) \) would block \( g^* \). If \( g^*(n_j) = n_{j+1} \), then \( n_j \) prefers \( g^* \) to \( g_1 \), so \( n_{j+1} \) cannot prefer \( g^* \) to \( g_1 \), or \( (n_j, n_{j+1}) \in g^* \) would block \( g_1 \).
which contradicts our premise. Suppose Lemma A.2 analogous to Lemma 3.1.

Corollary A.1 states \( g \in C \) and \( g \notin OS \). If \( i \in S, k_j \in K, (i, k_j) \in N_L(g), then g(i) \neq \emptyset and/or g(k_j) \neq \emptyset.

Proof. Suppose \( g(i) = \emptyset \) and \( g(k_j) = \emptyset \). Then, in any \( g' \in C, g'(i) = \emptyset \) and there exists \( k_l \in K \) such that \( g'(k_l) = \emptyset \). As \( (i, k_j) \in N_L(g), g + ik_j \in L(g) \). Let \( g^* \in OS \subseteq C, k_l \in K \) such that \( g^*(k_l) = \emptyset \). Then \( c_L(g^*) \leq c(g^*, g^* + ik_l) = c(g, g + ik_j) = c_L(g) \). Therefore \( g \in OS \), which contradicts our premise.

The next lemma is analogous to Lemma 3.2.

\footnote{Such a member must be in a cycle since players who are not in cycles are indifferent between \( g_1 \) and \( g^* \).}
Lemma A.3  Suppose that \( g \in \mathcal{C}, g \notin OS, \) and Assumption 4 holds. Suppose that \((i, k_j) \in N_L(g)\). Let \( K, K' \in K \) be such that \( g(i) \in K \) and \( k_j \in K' \). Then, for all \( g^* \in OS, \) either \( g^*(i) \notin K \) and/or \( g(k_j) \notin g^*(K') \).

Proof. Let \( g^* \in OS \). Suppose \( g^*(i) \in K \), and \( g(k_j) = g^*(k_j) \) for some \( k_j \in K' \). If \( K = K' \), then by Assumption 4, \( i = g(k_j) \) and \( c_L(g^*) \leq c(g^*, g^* - i) = c(g, g - ik_j) = c_L(g) \). If \( K \neq K' \), then \( c_L(g^*) \leq c(g^*, g^* + ik_j) = c(g, g + ik_j) = c_L(g) \). Therefore \( g \in OS \), which contradicts our premise. \( \square \)

Lemma A.4 (Getting Closer Lemma II) Suppose the dynamic satisfies Assumptions 4, 5. Let \( g' \in OS \). Suppose that \( g \in \mathcal{C} \) and \( g \notin OS \). Let \( g_1 \in L(g) \). Then, \( \exists g'' \in \mathcal{C}, t \in \mathbb{N}_+, \) such that \( \tilde{m}(g', g'') > \tilde{m}(g', g) \) and \( P_0^T(g_1, g'') > 0 \).

Proof. Let \( g^* \) satisfy:

\[
g^* \in \text{argmax } m(g, \hat{g}) \quad \hat{g} \in EQ(g')
\]

and

\[
g^* \in \text{argmax } m(g_1, \hat{g}) \quad \hat{g} \in EQ(g')
\]

It is possible to choose such a \( g^* \) as any student matched to the same college in \( g \) and \( g_1 \) is matched to the same position at that college.

Suppose that \( g - ig(i) \in L(g) \). Suppose that \( i \in S \). Let \( g(i) \in K \) and \( g^*(i) \in K^* \). Under Assumption 5, \( g' \in OS \) implies \( g^* \in OS \). This, and \( g \notin OS \) imply \( g(i) \in K \neq K^* \cap g^*(i) \), so \( m(g^*, g_1) = m(g^*, g) \), and as \( \tilde{m}(g', g_1) = m(g^*, g_1) \) and \( \tilde{m}(g', g) = m(g^*, g) \), we have \( \tilde{m}(g', g_1) = \tilde{m}(g', g) \). Since \( g_1 \) is unstable (i is single), Lemma 5.5 guarantees there exists \( T \in \mathbb{N}_+, g_T \in G, \) such that \( P_0^T(g_1, g_T) > 0 \) and \( \tilde{m}(g', g_T) > \tilde{m}(g', g_1) = \tilde{m}(g', g) \).

Next, suppose that \( g + ik_j \in L(g) \). If \( g(i) \notin \emptyset \), let \( K \) be such that \( g(i) \in K \) and \( K^* \) be such that \( g^*(i) \in K^* \). Let \( k_j \in K_j \). Lemma A.3 implies that \( g(i) \notin \emptyset \) and/or \( K \neq K^* \) and/or \( \emptyset \neq g(k_j) \notin g^*(K_j) \). Furthermore, Lemma A.2 implies that \( g(i) \notin \emptyset \) and/or \( g(k_j) \notin \emptyset \).

Case I: \( g(i) \notin \emptyset, K \neq K^* \) and \( \emptyset \neq g(k_j) \notin g^*(K_j) \) and/or \( (ik_j \notin g^*) \).

Note \( \tilde{m}(g', g_1) = m(g^*, g_1) \geq m(g^*, g) = \tilde{m}(g', g) \), with the inequality strict if \( ik_j \in g^* \). If \( ik_j \notin g^*, \) since \( g_1 \) is unstable \( (g(k_j) \) is single), Lemma 5.5 implies that there exists \( T \in \mathbb{N}_+, g_T \in G, \) such that \( P_0^T(g_1, g_T) > 0 \) and \( \tilde{m}(g', g_T) > \tilde{m}(g', g_1) = \tilde{m}(g', g) \).

Case II: \( g(i) \notin \emptyset, K \neq K^* \) and \( \emptyset \neq g(k_j) \notin g^*(K_j) \) or \( g(k_j) = \emptyset \) and/or \( (ik_j \notin g^*) \).
By definition of $g^*$, it must be that $g(k_j) = g^*(k_j)$. Note that $m(g^*, g_1) \geq m(g^*, g) - 2 \geq 0$ as $g_1(k_j) \neq g(k_j) = g^*(k_j)$.

If $g^*(g(i)) = \emptyset$, then $g(K) = g^*(K) \neq 0$, so $g^*(i) \in K$ and we have a contradiction. Therefore $g^*(g(i)) \neq \emptyset$.

First, suppose that $g^*(g(i))$ is indifferent between $g$ and $g^*$. This implies $g^*(g(i))$ and $g^*(g^*(g(i))) = g(i)$ are positions in the same college. By definition of $g^*$ and lemma 5.6 we have $g(g^*(g(i))) = g^*(g^*(g(i))) = g(i)$. So $g^*(g(i)) = i$, implying in turn that $g(i) = g^*(i)$ and $K = K^*$ which is a contradiction. Therefore $g^*(g(i))$ is not indifferent between $g$ and $g^*$.

Second, suppose that $g^*(g(i))$ prefers $g^*$ to $g$. Let $g_2 = g_1 + g(i)g^*(g(i))$. Recall that $g(i)$ is single, that $g^*(g(i))$ is either single or indifferent between $g$ and $g_1$, and that Assumption 4 does not prevent $g(i)$ and $g^*(g(i))$ from being matched, so $P_0(g_1, g_2) > 0$. Note that $m(g', g_2) \geq m(g^*, g_2) \geq m(g^*, g) = m(g', g)$. It cannot be that $g^*(g(i)) = g(k_j)$ as by $g(k_j) = g^*(k_j)$ we then have that $g^*(g(i)) = g^*(g(i))$ which would imply $g(i) = k_j$, contradicting $g + ik_j \in L^0(g)$. If $g(k_j) \neq \emptyset$, then $i \in K$ implies $i$ is not in any stable matching and $g_2$ is unstable. Lemma 5.5 implies that there exists $T \in \mathbb{N}_+, g_T \in G$, such that $P_0^T(g_2, g_T) > 0$ and $m(g', g_T) > m(g', g)$. Therefore $g^*(i)$ prefers $g^*$ to $g_1$. Let $i$ and $g^*(i)$ get matched. Let $g_4$ denote the resulting network. Note that $m(g', g_4) \geq m(g^*, g_4) \geq m(g^*, g) = m(g', g)$. $g(i)$ is single and $g^*(i) \notin K$, so $|g_4(K)| < |g'(K)|$ and $g_4$ is unstable. Lemma 5.5 implies that there exists $T \in \mathbb{N}_+, g_T \in G$, such that $P_0^T(g_4, g_T) > 0$ and $m(g', g_T) > m(g', g_4) \geq m(g', g)$.

\[21\] Roth (1986) tells us that any college with unfilled places in some stable matching is matched to the same set of students in any stable matching. A corollary of this is that any college must be matched to the same number of students in any stable matching.

\[22\] They strictly do so. $g(i)$ and $g^*(i)$ are college positions and have strict preferences over students. $K \neq K^*$ implies that student $i$ has strict preferences over $g$ and $g^*$.

\[23\] If $g(k_j) \neq \emptyset$ then $g^*(k_j)$ is single in $g_1$ because $g(k_j) = g^*(k_j)$. Assumption 4 does not prevent $k_j$ and $g^*(k_j)$ from being matched.

\[24\] $ik_j \notin g^*$ and the definition of $g^*$ imply that $i$ is matched to different colleges in $g_1$ and $g^*$. 

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Case III: \( g(i) = \emptyset \) or \( g(i) \neq \emptyset, K = K^* \) and \( \emptyset \neq g(k_j) \notin g^*(K_j) \) and \( (i_k_j \notin g^*) \).

Note that \( m(g^*, g_1) \geq m(g^*, g) - 2 \) as \( g(k_j) \notin g^*(K_j) \). First, suppose that \( i \) prefers \( g \) to \( g_1 \). If \( g(i) \neq \emptyset \), let \( i \) and \( g(i) \) get matched. If \( g(i) = \emptyset \), let \( i \) leave \( k_j \) to be single. Let \( g_5 \) denote the resulting network. \( g(i) \in K = K^* \) or \( g(i) = g^*(i) = \emptyset \) implies that \( m(g^*, g_5) \geq m(g^*, g) \). Since \( g_5 \) is unstable \( (g(k_j) \) is single), Lemma 5.5 implies that \( m(g^*, g_T) > m(g^*, g) \) with \( P_0^T(g_5, g_T) > 0 \) for some \( T \in \mathbb{N}^+, g_T \in G \).

Next, suppose that \( i \) prefers \( g_1 \) to \( g \). Then \( k_j \) prefers \( g \) and \( g^* \) to \( g_1 \). If \( k_j \) prefers \( g^* \) to \( g \), then \( g(k_j) \) prefers \( g \) to \( g^* \). This implies that \( g^*(g(k_j)) \) prefers \( g^* \) to \( g \). Let \( g^*(g(k_j)) \) and \( g(k_j) \) get matched. Let \( g_6 \) denote the resulting network. If \( g(g^*(g(k_j))) = \emptyset \), then by definition of \( g^* \) and Roth (1986) we have \( g^*(g(k_j)) = g(k_j) = \emptyset \), contradicting our assumptions for Case III. If \( g(g^*(g(k_j))) \neq \emptyset \), then \( g(g^*(g(k_j))) \) is single in \( g_6 \), so \( g_6 \) is unstable. Lemma 5.5 implies that \( m(g^*, g_T) > m(g^*, g_6) \geq m(g^*, g) \) with \( P_0^T(g_6, g_T) > 0 \) for some \( T \in \mathbb{N}^+, g_T \in G \).

If \( k_j \) prefers \( g \) to \( g^* \), then \( g^*(k_j) \) prefers \( g^* \) to \( g \). Let \( k_j \) and \( g^*(k_j) \) get matched. Let \( g_7 \) denote the resulting network. Note that \( m(g^*, g_7) \geq m(g^*, g_7) \geq m(g^*, g) = m(g', g) \). Since \( g_7 \) is unstable \((g(k_j) \) is single), Lemma 5.5 implies that \( m(g^*, g_T) > m(g^*, g_7) \geq m(g^*, g) \) with \( P_0^T(g_7, g_T) > 0 \) for some \( T \in \mathbb{N}^+, g_T \in G \).

For all cases, we have shown that there exists \( T \in \mathbb{N}^+, g_T \in G \), such that \( P_0^T(g_1, g_T) > 0 \) and \( m(g', g_T) > m(g', g) \). If \( g_T \in \mathcal{C} \), then we are done by letting \( g_T = g'' \). If \( g_T \notin \mathcal{C} \), then repeated application of Lemma 5.5 will lead the process to \( g'' \in \mathcal{C} \) such that \( m(g^*, g'') > m(g', g) \).  

**Proof of Theorem 5.8.** If \( g \in SS \), then \( g \in \mathcal{C} \) and there exists a minimal cost spanning tree rooted at \( g \). Denote the cost of this tree by \( \text{cost}(g) \). Assume \( g \notin OS \). Choose \( g' \in OS \). Construct a path of edges \( (g = g^1, \ldots, g^L) \) such that \( g^i \in \mathcal{C} \), \( g^i \notin OS \) for \( i = 1, \ldots, L - 1 \), and \( g^L \in OS \). The path is constructed as follows. For each \( g^i, i = 1, \ldots, L - 1 \), Lemma A.4 implies:

\[
\exists g^{i+1} \in \mathcal{C} : \quad m(g^i, g^{i+1}) > m(g^i, g^i) \quad \text{and} \quad C(g^i, g^{i+1}) = c_L(g^i).
\]

This is repeated until we reach some \( g^L \in OS \). Add these edges to the conjectured minimal cost spanning tree, replacing the existing edges exiting \( g^2, \ldots, g^{L-1} \). Remove the edge
leaving $g^L$. Denote the cost of the new tree by $\text{cost}(g^L)$. Then:

$$\text{cost}(g^L) \leq \text{cost}(g) + c_L(g) - c_L(g^L) < \text{cost}(g).$$

The first inequality follows from the construction of the tree rooted at $g^L$; the second inequality holds as $g \notin OS$ implies $c_L(g) < c_L(g^L)$. So, the conjectured minimal cost spanning tree can have been no such thing. Contradiction. 

References


