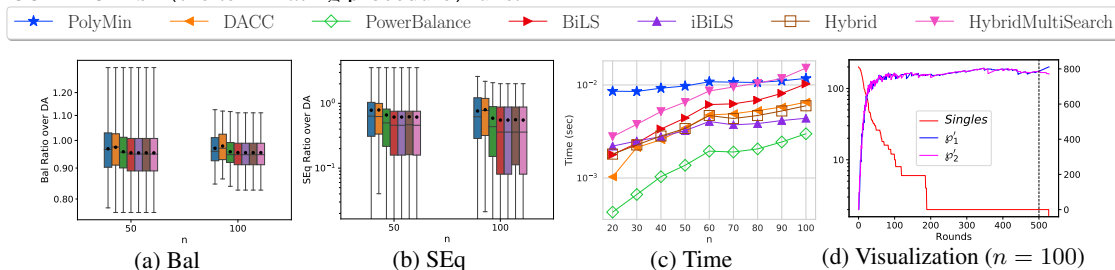


1 We thank all reviewers for the time they invested to review this paper and share their insights. In this letter, we respond
 2 to all reviewer comments, quoted verbatim, **bold in teal color**; content from the paper is quoted in blue.

3 **Application of the proposed algorithm in practice; visualization of the matching process; real-world stable**
 4 **marriage data.** We have conducted experiments on real-world data, yet could not include them within page limits.
 5 We extract distributions from the data of an online dating service [2]. The data consists of 17,359,346 anonymous
 6 ratings, on the 1 – 10 scale, of 168,791 profiles made by 135,359 LibimSeTi users, along with gender information. We
 7 remove users of unknown gender and those who have not rated the opposite gender, and construct a 2D distribution
 8 of the frequency of each pair of ratings (i, j) . Drawing from this distribution, we generate data of $n = 100$; the
 9 limited scale of ratings does not suffice to generate interesting preference lists at larger size. We resolve ties using 80%
 10 randomness and 20% popularity (P), i.e., the global ranking of agents by all ratings. We run 50 instances per size, and
 11 plot quality and runtime results. We visualize the process for POWERBALANCE with the instance that yields the *median*
 12 Sex-Equality Cost. In each round, we measure the number of *Single* agents and the sum of κ index values from men
 13 (ϕ'_m) and women (ϕ'_w), which dictate which side proposes in the following round. The left-side axis marks the scale of
 14 singles and the right-side axis marks the scale of ϕ'_m and women ϕ'_w . The vertical black dashed line marks the round in
 15 which COMPROMISE (the terminating procedure) runs.



16 **Publication of the algorithm in an implemented code (e.g. Java as stated in Line 304).** Please note that a link to
 17 code and data is provided in footnote 3 at Line 304, Page 7.

18 **Pseudocode for the subroutines PROPOSE and COMPROMISE.** The pseudocodes are given below.

<pre> procedure PROPOSE(p, μ) if $(\mu(p) = \emptyset \wedge \kappa_p < n)$ then $q = \ell_p[\kappa_p]$ if accept(q, p) then if $\mu(q) \neq \emptyset$ then $r = \mu(q); \mu = \mu \setminus \{(q, r)\}$ $\mu = \mu \cup \{(p, q)\}; \kappa_q = pr_q(p)$ else $\kappa_p = \kappa_p + 1$ </pre>	<p>▷ proposer p, matching μ</p> <p>▷ p wants to propose to q</p> <p>▷ break up q if married</p> <p>▷ match p and q</p> <p>▷ q rejects p</p>	<pre> function COMPROMISE(\mathcal{C}, μ) if $(\mathcal{C} = \mathcal{M})$ then $\mathcal{F} = \mathcal{W}$ else $\mathcal{F} = \mathcal{M}$ while $(\exists x \in \mathcal{C} : \mu(x) = \emptyset \wedge \kappa_x < n)$ do for all $x \in \mathcal{C}$ do PROPOSE(x, μ) while $(\exists x \in \mathcal{F} : \mu(x) = \emptyset \wedge \kappa_x < n)$ do for all $x \in \mathcal{F}$ do PROPOSE(x, μ) return μ </pre>	<p>▷ side \mathcal{C}, matching μ</p>
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19 **Relevance to real world matching problems unclear.** As we report in Lines 33–34, citing recent results by Hassidim
 20 et al. [3] brought to our attention by Roth [4], the set of possible stable matchings is large in real-world markets [3].

21 **What does “SEQ ratio over DA” mean?** It means that we normalize cost results, dividing by the corresponding
 22 best cost the DA algorithm can obtain; lower cost values are better.

23 **A characterization of under what cases DA leads to very unfair matchings.** The most challenging cases are those
 24 of *symmetric* distributions on two sides, even more so when some choices are universally popular, as in dataset D.

25 **Minor comments regarding language/notation/citations/etc.** Thank you for these comments, we will heed them.

26 **“10 million students” in a Chinese admissions market.** We cite this fact as a motivation for research. In most
 27 countries the sizes of student admission markets are in the order of a few thousands.

28 **Why not compare against an IP formulation of the problem?** A mixed-integer linear programming formulation is
 29 indeed possible, e.g., minimizing an auxiliary variable X such that $SEq < X$ and $SEq > -X$. To our knowledge,
 30 such an IP-based solution has not been attempted to date. Sethuraman et al. [6] consider fairness, yet only in terms
 31 of a *median* stable matching, not in terms of sex-equality cost. Sethuraman opined that an IP-based solution method
 32 for sex-equal stable marriage is “unlikely to have polynomial running time” [5]. In a recent study [1], Ágoston et al.
 33 were able to solve only a pruning-intensive variant of the stable matching problem using an IP technique on real data;
 34 other variants were infeasible for sizes larger than 100; matching 100 students to 20 schools with common quotas took
 35 39,560 seconds (almost 11 hours) [1]. On the other hand, our quadratic-time algorithms surpass or match the quality
 36 that APPROX achieves with all carefully tested values of ε for which a solution exists (Figure 3); thus, they achieve
 38 *near-optimal* solutions. We reconfirmed this fact in communication with the authors of APPROX [7].

39 [1] Kolos Csaba Ágoston, Péter Biró, and Iain McBride. Integer programming methods for special college admissions problems. *Jnl Comb. Opt.*, 32(4):, Nov 2016. 1

40 [2] Lukas Brozovsky and Vaclav Petricek. Recommender system for online dating service. In *Proceedings of Conference Znalosti 2007*, Ostrava, 2007. VSB. 1

41 [3] Avinatan Hassidim, Assaf Romm, and Ran I. Shorrer. Need vs. merit: The large core of college admissions markets. <https://ssrn.com/abstract=3071873>. 1

42 [4] Alvin E. Roth. Private Communication, 2018. 1

43 [5] Jay Sethuraman. Private Communication, 2019. 1

44 [6] Jay Sethuraman, Chung-Piaw Teo, and Liwen Qian. Many-to-one stable matching: Geometry and fairness. *Math. Operations Research*, 31(3):581–596, 2006. 1

45 [7] Hiroki Yanagisawa and Shuichi Miyazaki. Private Communication, 2019. 1