Redesign of the National Resident Matching Program

We turn next to think about

• Theory as an input to design
• and why it’s not the only input,
• and what some other inputs might be...

To put it another way, we want to think about what makes design difficult, taking as a case study the redesign of the NRMP.

Some NRMP "match variations:" What makes the NRMP different from a simple college admissions model is that it has complications which sometimes cause two positions to be linked to one another, and sometimes cause the number of positions to change.

In the first category of differences are *couples*, who submit rank orders of pairs of programs and must be matched to two positions; and applicants who match to 2nd year positions and have *supplemental lists* which must then be consulted to match them to 1st year positions.

In the second category are requests by residency programs to have an *even or an odd number of matches*, and *reversions of unfilled positions from one program to another.*
These complications matter for two related reasons:

1. They may change the properties of the match; and
2. The clearinghouse algorithm must be designed to accommodate them

Let’s take a look at how we might model couples, for example (keeping in mind that we’ll eventually have to take account of all the match variations, not just couples...)
Married couples looking for two residencies weren’t an issue in the medical job market until the 1970’s

Table 13.—Women in US Medical Schools

<table>
<thead>
<tr>
<th>Academic Year*</th>
<th>Women Applicants†</th>
<th>Women in Entering Class</th>
<th>Total Women Enrolled</th>
<th>Graduates</th>
</tr>
</thead>
<tbody>
<tr>
<td>1967-1968</td>
<td>1951 (10.4)</td>
<td>934 (9.9)</td>
<td>3003 (8.7)</td>
<td>641 (8.0)</td>
</tr>
<tr>
<td>1977-1978</td>
<td>10195 (25.1)</td>
<td>4149 (25.7)</td>
<td>14373 (23.8)</td>
<td>3086 (21.4)</td>
</tr>
<tr>
<td>1980-1981</td>
<td>10644 (29.5)</td>
<td>4970 (28.9)</td>
<td>17373 (26.5)</td>
<td>3892 (24.8)</td>
</tr>
<tr>
<td>1981-1982</td>
<td>11673 (31.8)</td>
<td>5343 (30.8)</td>
<td>18555 (27.9)</td>
<td>3991 (25.0)</td>
</tr>
<tr>
<td>1982-1983</td>
<td>11685 (32.7)</td>
<td>5445 (31.6)</td>
<td>19627 (29.3)</td>
<td>4229 (26.7)</td>
</tr>
<tr>
<td>1983-1984</td>
<td>11961 (33.9)</td>
<td>5659 (32.9)</td>
<td>20685 (30.7)</td>
<td>4617 (28.3)</td>
</tr>
<tr>
<td>1984-1985</td>
<td>12476 (34.7)</td>
<td>5705 (33.8)</td>
<td>21287 (31.7)</td>
<td>4898 (30.0)</td>
</tr>
<tr>
<td>1985-1986</td>
<td>11562 (35.1)</td>
<td>5788 (34.2)</td>
<td>21624 (32.5)</td>
<td>4930 (30.8)</td>
</tr>
<tr>
<td>1986-1987</td>
<td>11267 (36.0)</td>
<td>5886 (35.0)</td>
<td>22082 (33.4)</td>
<td>5092 (32.1)</td>
</tr>
<tr>
<td>1987-1988</td>
<td>10411 (37.0)</td>
<td>6087 (36.5)</td>
<td>22539 (34.3)</td>
<td>5219 (32.7)†</td>
</tr>
</tbody>
</table>

*Ponce (Puerto Rico) School of Medicine and the University of South Dakota, Sioux Falls, did not provide information for the 1980-1981 academic year, so 1979-1980 enrollment data were used for these schools. Similarly, Howard University, Washington, DC, did not provide information for the 1987-1988 academic year, so 1986-1987 enrollment data were used for this school.
†Source of data: Medical School Admission Requirements, Association of American Medical Colleges, Section for Student Services.
‡Data estimated in April 1988.

An initial “couples algorithm” in the 1970’s

- Couples (after being certified by their dean) could register for the match as a couple.
  - They had to specify one member of the couple as the “leading member.”
  - They submitted a separate rank order list of positions for each member of the couple.
- The leading member went through the match as if single.
- The other member then had his/her rank order list edited to remove positions not in the ‘same community’ as the one the leading member had matched to.
  - Initially the NRMP determined communities; in a later version, when couples were still defecting, couples could specify this themselves.
A similar algorithm is used in Scotland today

• “Linked applicants
• To accommodate linked applicants, a joint preference list is formed for each such pair, using their individual preference lists and the programme compatibility information. If such a pair, $a$ and $b$, have individual preferences $p_1, p_2, \ldots, p_{10}$ and $q_1, q_2, \ldots, q_{10}$ respectively (with $a$ the higher scoring applicant), then the joint preference list of the pair $(a,b)$ is $(p_1,q_1), (p_1,q_2), (p_2,q_1), (p_2,q_2), (p_1,q_3), (p_3,q_1), (p_2,q_3), (p_3,q_2), \ldots, (p_9,q_{10}), (p_{10},q_9), (p_{10},q_{10})$ (except that incompatible pairs of programmes are omitted) “
• http://www.nes.scot.nhs.uk/sfas/About/default.asp
But this didn’t work well for couples

• Why?
• The iron law of marriage: You can’t be happier than your spouse.
• Couples consume *pairs* of jobs. So an algorithm that only asks for their preference orderings over *individual* jobs can’t hope to avoid instabilities (appropriately redefined to include couples’ preferences)
• But even if we ask couples for their preferences over pairs of jobs, we may still have a problem: Roth (1984) observed that the set of stable matchings may be *empty* when couples are present.
A More Complex Market: Matching with Couples
This model is the same as the college admissions model, except the set of workers is replaced by a set of applicants that includes individuals and couples.

Denote the set of applicants by $A = A_1 \cup C$, where $A_1$ is the set of (single) applicants who seek no more than one position, and $C$ is the set of couples $\{a_i, a_j\}$ such that $a_i$ is in the set $A_2$ (of husbands) and $a_j$ is in the set $A_3$, and the sets of applicants $A_1$, $A_2$, and $A_3$ together make up the entire population of individual applicants, $A' = A_1 \cup A_2 \cup A_3$.

Each couple $c = \{a_i, a_j\}$ in $C$ has preferences over ordered pairs of positions, i.e. an ordered list of elements of $F \times F$. The first element of this list is some $(r_i, r_j)$ in $F \times F$ which is the couples' first choice pair of jobs for $a_i$ and $a_j$ respectively, and so forth.

Applicants in the set $A_1$ have preferences over residency programs, and residency programs (firms) have preferences over the individuals in $A'$, just as in the simple model discussed earlier. (That is, firms view the members of a couple as two distinct individuals...)
A matching is a set of pairs in $FxA$.

Each single applicant, each couple, and each residency program submits to the centralized clearinghouse a Rank Order List (ROL) that is their stated preference ordering of acceptable alternatives.

A matching $\mu$ is blocked by a single applicant (in the set $A1$), or by a residency program, if $\mu$ matches that agent to some individual or residency program not on its ROL.

A matching is blocked by an individual couple $(a_i,a_j)$ if they are matched to a pair $(r_i,r_j)$ not on their ROL.

A residency program $r$ and a single applicant $a$ in $A1$ together block a matching $\mu$ precisely as in the college admissions market, if they are not matched to one another and would both prefer to be.
A couple \( c=(a_1,a_2) \) and residency programs \( r_1 \) and \( r_2 \) block \( \mu \) if the couple prefers \( (r_1,r_2) \) to \( \mu(c) \), and either \( r_1 \) and \( r_2 \) each would prefer to be matched to the corresponding couple member, or if one of them would prefer, and the other already is matched to the corresponding couple member. That is, \( c \) and \( (r_1,r_2) \) block \( \mu \) if

1. \((r_1,r_2) >_c \mu(c)\); and if either
2. \(\{ a_1 \not\in \mu(r_1), \text{ and } a_1 >_{r_1} a_i \text{ for some } a_i \in \mu(r_1) \text{ or } a_1 \text{ is acceptable to } \mu(r_1) \} \) and either \( a_2 \in \mu(r_2) \) or \( \{ a_2 \not\in \mu(r_2), a_2 >_{r_2} a_j \text{ for some } a_j \in \mu(r_2) \text{ or } a_2 \text{ is acceptable to } \mu(r_2) \} \)
   or
3. \( a_1 \in \mu(r_1) \) and \( \{ a_2 \not\in \mu(r_2), a_2 >_{r_2} a_j \text{ for some } a_j \in \mu(r_2) \text{ or } a_2 \text{ is acceptable to } \mu(r_2) \} \)
A matching is \textit{stable} if it is not blocked by any individual or by a couple, together with one or two residency programs.

Theorem 5.11 (Roth ’84): In the college admissions model with couples, the set of stable matchings may be empty.

Proof: by (counter)example. (Which was sufficient, when I didn’t have design responsibility...😊)
Example--market with one couple and no stable matchings (motivated by Klaus and Klijn, and Nakamura (JET corrigendum 2009 to K&K JET 2005):

Let \( c=(s1,s2) \) be a couple, and suppose there is another single student \( s3 \), and two hospitals \( h1 \) and \( h2 \). Suppose that the acceptable matches for each agent, in order of preference, are given by

\[
\begin{align*}
\text{c: } & (h1,h2); & \text{s3: } & h1, h2, \\
\text{h1: } & s1, s3; & \text{h2: } & s3, s2
\end{align*}
\]

Then no individually rational matching \( \mu \) (i.e. no \( \mu \) that matches agents only to acceptable mates) is stable. We consider two cases, depending on whether the couple is matched or unmatched.

Case 1: \( \mu(c)=(h1,h2) \). Then s3 is unmatched, and s/he and h2 can block \( \mu \), because h2 prefers s3 to \( \mu(h2)=s2 \).

Case 2: \( \mu(c)=c \) (unmatched). If \( \mu(s3)=h1 \), then \( (c, h1,h2) \) blocks \( \mu \). If \( \mu(s3)=h2 \) or \( \mu(s3)=s3 \) (unmatched), then \( (s3,h1) \) blocks \( \mu \).
Why is the couples problem hard?

• Note first that the ordinary deferred acceptance algorithm won’t in general produce a stable matching (even when one exists, and even when couples state preferences over pairs of positions)
  – In the worker proposing algorithm, if my wife and I apply to a pair of firms in Boston, and our offers are held, and I am later displaced by another worker, my wife will want to withdraw from the position in which she is being held (and the firm will regret having rejected other applications to hold hers)
  – In the firm proposing algorithm, it may be hard for a couple to determine which offers to hold.
Furthermore, the following example shows that even when the set of stable matchings is non-empty, it may no longer have the nice properties we’ve come to expect.

Matching with couples (Example of Aldershof and Carducci, ’96)
4 hospitals \{h1,...,h4\} each with one position;
2 couples \{s1,s2\} and \{s3,s4\}

Preferences:

<table>
<thead>
<tr>
<th></th>
<th>h1</th>
<th>h2</th>
<th>h3</th>
<th>h4</th>
<th>{s1,s2}</th>
<th>{s3,s4}</th>
</tr>
</thead>
<tbody>
<tr>
<td>S4</td>
<td>s2</td>
<td>s2</td>
<td>s2</td>
<td>s2</td>
<td>h3h2</td>
<td>h2h1</td>
</tr>
<tr>
<td>S3</td>
<td>s3</td>
<td>s4</td>
<td>s3</td>
<td>s3</td>
<td>h2h3</td>
<td>h2h3</td>
</tr>
<tr>
<td>S1</td>
<td>s1</td>
<td>s1</td>
<td></td>
<td></td>
<td>h2h4</td>
<td>h1h3</td>
</tr>
</tbody>
</table>

h3h4          h4h1
h3u           h4h3
h2u           h2u
There are exactly two stable matchings: $h_1,...,h_4$ are either matched to:

\[
\begin{array}{cccc}
h_1 & h_2 & h_3 & h_4 \\
(s_4 & s_2 & s_1 & s_3) & h_1, h_2, h_4, \{s_1,s_2\} \text{ prefer this}
\end{array}
\]

or to

\[
\begin{array}{cccc}
(s_4 & s_3 & s_2 & u) & h_3, \{s_3,s_4\} \text{ prefer this}
\end{array}
\]

So, even when stable matchings exist, there need not be an optimal stable matching for either side, and employment levels may vary.
So we can start to note theorems about simple markets whose conclusions do not carry over to markets with the NRMP match variations (even just with couples).

In a simple matching market:

1. the set of stable matchings is always nonempty

2. the set of stable matchings always contains a "program optimal" stable matching, and an "applicant optimal" stable matching.

3. the same applicants are matched and the same positions are filled at every stable matching.

Similarly, strategic results about simple markets won’t carry over unchanged to the more complex medical market.
Strategic behavior in simple markets (without match variations):

1. In simple markets, *when the applicant proposing algorithm is used*, but not when the hospital proposing algorithm is used, *no applicant can possibly improve his match by submitting an ROL that is different from his true preferences*.

2. In simple markets when the program proposing algorithm is used, every applicant who can do better than to submit his true preferences as his ROL can do so by submitting a truncation of his true preferences.

3. In simple markets, when the program proposing algorithm is used, *the only applicants who can do better than to submit their true preferences are those who would have received a different match from the applicant proposing algorithm*.

Furthermore, the best such applicants can do is to obtain the applicant optimal match.
### Descriptive Statistics: NRMP

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<tr>
<td><strong>APPLICANTS</strong></td>
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<tr>
<td>Primary ROL’s</td>
<td>20071</td>
<td>20916</td>
<td>22353</td>
<td>22937</td>
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<tr>
<td>Applicants with Supplemental ROL’s</td>
<td>1572</td>
<td>2515</td>
<td>2312</td>
<td>2098</td>
<td>2436</td>
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<tr>
<td>Couples</td>
<td></td>
<td></td>
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<td></td>
<td></td>
</tr>
<tr>
<td>Applicants who are Coupled</td>
<td>694</td>
<td>854</td>
<td>892</td>
<td>998</td>
<td>1008</td>
</tr>
<tr>
<td><strong>PROGRAMS</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Active Programs</td>
<td>3225</td>
<td>3677</td>
<td>3715</td>
<td>3800</td>
<td>3830</td>
</tr>
<tr>
<td>Active Programs with ROL Returned</td>
<td>3170</td>
<td>3622</td>
<td>3662</td>
<td>3745</td>
<td>3758</td>
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<tr>
<td>Potential Reversions of Unfilled Positions</td>
<td></td>
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<tr>
<td>Programs Specifying Reversion</td>
<td>69</td>
<td>247</td>
<td>276</td>
<td>285</td>
<td>282</td>
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<tr>
<td>Positions to be Reverted if Unfilled</td>
<td>225</td>
<td>1329</td>
<td>1467</td>
<td>1291</td>
<td>1272</td>
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<tr>
<td>Programs Requesting Even Matching</td>
<td>4</td>
<td>2</td>
<td>6</td>
<td>7</td>
<td>8</td>
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<tr>
<td>Total Quota Before Match</td>
<td>19973</td>
<td>22737</td>
<td>22801</td>
<td>22806</td>
<td>22578</td>
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</tbody>
</table>
Because conclusions about simple markets can have counterexamples in the complex medical market, there were points at which we had to rely on computational explorations to see how close an approximation the simple theory provides for the complex market.

We relied on computation in three places:

Computational experiments were used in the algorithm design.

Computational explorations of the data from previous years were used to study the effect of different algorithms.

Theoretical computation, on simple markets, was used to understand the relation between market complexity and market size.
Before we talk about computational experiments, let’s think about what class of algorithms we might want to explore.

The deferred acceptance algorithm for simple matching models gave us a one pass algorithm; it starts with everyone unmatched, and never cycles. Our sense was that we weren’t going to be able to do that in the complex problem, but rather that we’d have to build a stable matching (if one existed), resolving blocking pairs as we identified them.

E.g. if \((m',w')\) is a blocking pair for a matching \(\mu\), a new matching \(\nu\) can be obtained from \(\mu\) by *satisfying* the blocking pair if \(m'\) and \(w'\) are matched to one another at \(\nu\), their mates at \(\mu\) (if any) are unmatched at \(\nu\), and all other agents are matched to the same mates at \(\nu\) as they were at \(\mu\).

But even for the simplest models (e.g. the marriage model) this can cycle, and might not converge to a stable matching (an observation originally due to Donald Knuth: see the discussion of Example 2.4 in Roth and Sotomayor).

In the marriage model there’s a way around this problem.
**Theorem 2.33** (Roth and Vande Vate): Let $\mu$ be an arbitrary matching for $(M, W, P)$. Then there exists a finite sequence of matchings $\mu_1, \ldots, \mu_k$, such that $\mu = \mu_1, \ldots \mu_k$ is stable, and for each $i = 1, \ldots, k-1$, there is a blocking pair $(m_i, w_i)$ for $\mu_i$ such that $\mu_{i+1}$ follows from $\mu_i$ by satisfying the pair $(m_i, w_i)$. 
Elements of the proof: Let $\mu_1$ be an arbitrary (w.l.o.g individually rational) matching with blocking pair $(m_1,w_1)$. Let $\mu_2$ be the matching obtained by satisfying the blocking pair, and define the set $A(1) = \{m_1,w_1\}$.

**Inductive assumption:** Let $A(q)$ be a subset of $M \cup W$ such that there are no blocking pairs for $\mu_{q+1}$ contained in $A(q)$, and such that $\mu_{q+1}$ does not match any agent in $A(q)$ to any agent outside of $A(q)$.

Then if $\mu_{q+1}$ isn’t stable, there is a blocking pair $(m’,w’)$ such that at most one of $m’$ and $w’$ is contained in $A(q)$. (If neither of $\{m’,w’\}$ is in $A(q)$, let $A(q+1) = A(q) \cup \{m’,w’\}$ and let $\mu_{q+2}$ be obtained from $\mu_{q+1}$ by satisfying the blocking pair $(m’,w’)$.

Otherwise, one of the pair is in $A(q)$, say $m’$ (in the other case the symmetric argument will apply). Let $A(q+1) = A(q) \cup \{w’\}$. Now run the deferred acceptance algorithm, just in the set $A(q+1)$, starting with $w’$ proposing and continuing until a matching is reached with no blocking pairs among the members of $A(q+1)$. The output is $\mu_{q+2}$.
This suggests a new class of algorithms, of which the deferred acceptance algorithm is a special case.

Start with an arbitrary matching \( \mu \), and select a subset \( A \) of agents such that there are no blocking pairs for \( \mu \) contained in \( A \), and \( \mu \) does not match any agent in \( A \) to any agent not in \( A \).

(For example, \( A \) could be a pair of agents matched under \( \mu \), or a single agent, or the set of all men.)

A new player, say woman \( w \), is selected to join \( A \). If no man in \( A \) is part of a blocking pair with woman \( w \), we may simply add her to \( A \) without changing the matching. Otherwise, select the man \( m \) whom woman \( w \) most prefers among those in \( A \) with whom she forms a blocking pair, and form a new matching by satisfying this blocking pair. If there is a woman \( w' = \mu(m) \), then she is left unmatched at this new matching, and so there may now be a blocking pair \((w', m')\) contained in \( A \). If so, choose the blocking pair most preferred by \( w' \) to form the next new matching.
The process continues in this way within the set $A \cup \{w\}$, like the deferred acceptance algorithm with women proposing, satisfying the blocking pairs which arise at each step until the process terminates with a matching $\mu_i$ having no blocking pairs within $A_i = S \cup \{w\}$.

The process can now be continued, with the selected set $A_i$ growing at each stage. At each stage, the selected set has no blocking pairs in it for the associated matching $\mu_i$, and so the process converges to a stable matching when $A_k = M \cup W$.

In the deferred acceptance algorithm with men proposing, the initial matching $\mu$ is the one at which all agents are single, and the initial set $A$ is $A=W$.

In the deferred acceptance algorithm with men proposing, the welfare of the women rises monotonically throughout the algorithm. In this more general class of algorithms there is no parallel, since agents from either side may be introduced into the set $A$. But the set $A$ itself grows, so the algorithm converges.

So, we’ll be looking for an algorithm that accumulates agents not involved in blocking pairs...
Computational experiments in the algorithm design.

The process by which the applicant-proposing algorithm was designed is roughly as follows. First, a conceptual design was formulated and circulated for comment, based on the family of algorithms explored (for the marriage model) in Roth and Vande Vate (1990). ("Random Paths to Stability in Two-Sided Matching," *Econometrica*, 58, 1990, 1475-1480.)

To code this into a working algorithm, a number of choices had to be made, concerning the sequence in which operations would be conducted.

Most of these decisions can be shown to have no effect on the outcome of *simple* matches, but could potentially effect the outcome when the NRMP match variations are present.

Consequently, we performed computational experiments before making sequencing choices.

(Throughout the algorithm design process, we also posted progress reports on the web, where they were available to all interested parties. Since designs often have to be adopted by multiple constituencies, the design *process* might be important...)
• Do sequencing differences cause substantial or predictable changes in the match result (e.g. do applicants or programs selected first do better or worse than their counterparts selected later)?

• Does the sequence of processing affect the likelihood that an algorithm will produce a stable matching?

Experiments to test the effect of sequencing were conducted using data from three NRMP matches: 1993, 1994, and 1995.

The results were that sequencing effects existed, but were unsystematic, and effected on the order of 1 in 10,000 matches.

(In the majority of years and algorithm sequences examined, the match was unaffected by changes in sequencing of algorithm operations, and in the majority of the remaining cases only 2 applicants received different matches.)

However sequencing decisions did influence the speed of convergence to a stable matching.
Based on the sequencing experiments described above, the following decisions were made pertaining to the design of the applicant proposing algorithm for the NRMP:

1. **All single applicants are admitted to the algorithm for processing before any couples are admitted.**

2. Single applicants are admitted for processing in ascending sequence by applicant code.

3. Couples are admitted for processing in ascending sequence by the lower of the two applicant codes of the couple.

When a program is selected from the program stack for processing, the applicants ranked by the program are processed in ascending order by program rank number (i.e. in order of the program’s preferences).
Computational Exploration of the Difference Between Program and Applicant Proposing Algorithms

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<tbody>
<tr>
<td><strong>APPLICANTS</strong></td>
<td></td>
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<td></td>
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<tr>
<td>Number of Applicants Affected</td>
<td>20</td>
<td>16</td>
<td>20</td>
<td>14</td>
<td>21</td>
</tr>
<tr>
<td>Applicant Proposing Result Preferred</td>
<td>12</td>
<td>16</td>
<td>11</td>
<td>14</td>
<td>12</td>
</tr>
<tr>
<td>Program Proposing Result Preferred</td>
<td>8</td>
<td>0</td>
<td>9</td>
<td>0</td>
<td>9</td>
</tr>
<tr>
<td>New Matched</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
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<tr>
<td>New Unmatched</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td><strong>PROGRAMS</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of Programs Affected</td>
<td>20</td>
<td>15</td>
<td>23</td>
<td>15</td>
<td>19</td>
</tr>
<tr>
<td>Applicant Proposing Result Preferred</td>
<td>8</td>
<td>0</td>
<td>12</td>
<td>1</td>
<td>10</td>
</tr>
<tr>
<td>Program Proposing Result Preferred</td>
<td>12</td>
<td>15</td>
<td>11</td>
<td>14</td>
<td>9</td>
</tr>
<tr>
<td>Programs with New Position(s) Filled</td>
<td>0</td>
<td>0</td>
<td>2</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>Programs with New Unfilled Positions</td>
<td>1</td>
<td>0</td>
<td>2</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>
If this were a simple market, the small number of applicants whose matching is changed when we switch from hospitals proposing to applicants proposing would imply that there was also little room for strategic behavior when it comes time to state rank order lists.

We can find out if this is also true in the complex market with computational experiments. It turns out that we don’t have to experiment on each individual separately, to put an upper bound on how many individuals could profitably manipulate their preferences.

(For the moment, we treat the submitted preferences as the true preferences—we’ll see in a minute why that is justified.)
Computational experiments to find upper bounds for the scope of strategic behavior

Truncations at the match point (to check if examining truncations is sufficient in the multi-pass algorithm...)

Difference in result for both the program proposing algorithm and the applicant proposing algorithm when applicant ROLs truncated at the match point:

<table>
<thead>
<tr>
<th></th>
<th>1993</th>
<th>1994</th>
<th>1995</th>
</tr>
</thead>
<tbody>
<tr>
<td>none</td>
<td>none</td>
<td>2 applicants improve, same positions filled</td>
<td>2 applicants improve, same positions filled</td>
</tr>
</tbody>
</table>

Difference in result for the program proposing algorithm when program ROLs truncated at the match point:

<table>
<thead>
<tr>
<th></th>
<th>1993</th>
<th>1994</th>
<th>1995</th>
</tr>
</thead>
<tbody>
<tr>
<td>none</td>
<td>none</td>
<td>none</td>
<td>2 applicants do worse, same positions filled</td>
</tr>
</tbody>
</table>
Difference in result for the applicant proposing algorithm when *program* ROLs truncated *at* the match point:

<table>
<thead>
<tr>
<th>1993</th>
<th>1994</th>
<th>1995</th>
</tr>
</thead>
<tbody>
<tr>
<td>none</td>
<td>3 applicants do worse, same number of positions filled, but not same positions [3 programs filled one less position, 1 program filled 1 more position, 1 program filled 2 more positions, 1 additional position was reverted from one program to another].</td>
<td>none</td>
</tr>
</tbody>
</table>
## Results for Iterative Truncations of Applicant ROL’s just *above* the match point

<table>
<thead>
<tr>
<th></th>
<th>1993</th>
<th></th>
<th>1994</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Program Proposing Algorithm</td>
<td>Applicant Proposing</td>
<td>Program Proposing Algorithm</td>
<td>Applicant Proposing</td>
</tr>
<tr>
<td></td>
<td>Truncated</td>
<td>Truncated &amp; Improved</td>
<td>Truncated</td>
<td>Truncated &amp; Improved</td>
</tr>
<tr>
<td>Run 1</td>
<td>17209</td>
<td>4546</td>
<td>17209</td>
<td>4536</td>
</tr>
<tr>
<td>Run 2</td>
<td>4546</td>
<td>2093</td>
<td>4536</td>
<td>2082</td>
</tr>
<tr>
<td>Run 3</td>
<td>2093</td>
<td>1036</td>
<td>2082</td>
<td>1023</td>
</tr>
<tr>
<td>Run 4</td>
<td>1036</td>
<td>514</td>
<td>1023</td>
<td>498</td>
</tr>
<tr>
<td>Run 5</td>
<td>514</td>
<td>258</td>
<td>498</td>
<td>241</td>
</tr>
<tr>
<td>Run 6</td>
<td>258</td>
<td>135</td>
<td>241</td>
<td>116</td>
</tr>
<tr>
<td>Run 7</td>
<td>135</td>
<td>73</td>
<td>116</td>
<td>52</td>
</tr>
<tr>
<td>Run 8</td>
<td>73</td>
<td>48</td>
<td>52</td>
<td>25</td>
</tr>
<tr>
<td>Run 9</td>
<td>48</td>
<td>34</td>
<td>25</td>
<td>12</td>
</tr>
<tr>
<td>Run 10</td>
<td>34</td>
<td>27</td>
<td>12</td>
<td>5</td>
</tr>
<tr>
<td>Run 11</td>
<td>27</td>
<td>24</td>
<td>5</td>
<td>2</td>
</tr>
<tr>
<td>Run 12</td>
<td>24</td>
<td>22</td>
<td>2</td>
<td>0</td>
</tr>
<tr>
<td>Run 13</td>
<td>22</td>
<td>22</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
The truncation experiments with applicants' ROLs yield the following upper bounds for the two algorithms in the years studied.

Upper limit of the number of applicants who could benefit by truncating their lists at one above their original match point:

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Program-Proposing Algorithm</td>
<td>12</td>
<td>22</td>
<td>13</td>
<td>16</td>
<td>11</td>
</tr>
<tr>
<td>Applicant-Proposing Algorithm</td>
<td>0</td>
<td>0</td>
<td>2</td>
<td>2</td>
<td>9</td>
</tr>
</tbody>
</table>

As expected, more applicants can benefit from list truncation under the program-proposing algorithm than under the applicant-proposing algorithm. Note that the number of applicants who could even potentially benefit from truncating their ROLs under the program-proposing algorithm is in each year almost exactly equal to the number of applicants who received a preferred match under the applicant proposing match (line 2 of Table 2). This suggests that this upper bound is very close to the precise number that would be predicted in the absence of match variations.
The truncation experiments with programs' ROLs yield the following upper bounds.

Upper limit of the number of programs that could benefit by truncating their lists at one above the original match point:

<table>
<thead>
<tr>
<th>Year</th>
<th>Program-Proposing Algorithm</th>
<th>Applicant-Proposing Algorithm</th>
</tr>
</thead>
<tbody>
<tr>
<td>1987</td>
<td>15</td>
<td>27</td>
</tr>
<tr>
<td>1993</td>
<td>12</td>
<td>28</td>
</tr>
<tr>
<td>1994</td>
<td>15</td>
<td>27</td>
</tr>
<tr>
<td>1995</td>
<td>23</td>
<td>36</td>
</tr>
<tr>
<td>1996</td>
<td>14</td>
<td>18</td>
</tr>
</tbody>
</table>

As expected, some programs can benefit from list truncation under either algorithm. However, consistently more programs benefit from list truncation under the applicant-proposing algorithm than under the program-proposing algorithm. Note that although the numbers of programs in these upper bounds remain small, they are in many cases about twice as large as the number of programs which received a preferred match at the stable matching produced by the algorithm other than the one being manipulated.

Refined estimate of the upper limit of the number of programs that could improve their results by truncating their own ROL’s in 1995 (Based on 50% sample):

<table>
<thead>
<tr>
<th></th>
<th>Program Proposing Algorithm</th>
<th>Applicant Proposing Algorithm</th>
</tr>
</thead>
<tbody>
<tr>
<td>Original Results</td>
<td>23</td>
<td>36</td>
</tr>
<tr>
<td>Current Estimate (still an upper limit)</td>
<td>12</td>
<td>22</td>
</tr>
</tbody>
</table>
Residency programs have another dimension on which they can manipulate; they not only have to report their preferences, but also how many positions they wish to fill. As we’ve seen in examples of the (simple) college-admissions model, and as in multi-unit auctions, they may potentially benefit from demand reductions.

(And for an impossibility theorem on avoiding capacity manipulation, see Sonmez, Tayfun [1997], "Manipulation via Capacities in Two-Sided Matching Markets," Journal of Economic Theory, 77, 1, November, 197-204.)

Revised Estimate of the Upper Bound of the Number of Programs That Could Improve Their Remaining Matches By Reducing Quotas

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Program Proposing Algorithm</td>
<td>28</td>
<td>16</td>
<td>32</td>
<td>8</td>
<td>44</td>
</tr>
<tr>
<td>Applicant Proposing Algorithm</td>
<td>8</td>
<td>24</td>
<td>16</td>
<td>16</td>
<td>32</td>
</tr>
</tbody>
</table>

This will be worth thinking about again—a small cloud on the horizon—when we consider what temptations may exist for residency programs to hire some of their people early, before the match. If there are such temptations, they may not be counterbalanced by a tendency to do worse in the match, on the contrary, reducing demand may have small spillover benefits in the match for the remaining candidates...
Overall, the striking thing about all these computational results is how small the set of stable outcomes appear to be; i.e. how few applicants or programs are affected by a switch from program proposing to applicant proposing, and how small are the opportunities to misrepresent preferences or capacities.

But we don’t really understand the structure of the set of stable matchings when there are couples, supplementary lists, and reversion of positions from one program to another. So there’s a chance that we’re making a big mistake here.

For example, we know that program and applicant optimal stable matchings no longer exist, but we’ve been studying the set of stable matchings by looking at the outcomes of the program and applicant proposing algorithms. Maybe the set of stable matchings isn’t all located between these two matchings when all the match variations are present; maybe the set of stable matchings just appears to be small because we don’t know where to look.

Even if our conclusions are correct, we’d like to know why. Could it be some spooky interaction between the size of the market and the presence of complications? Or does the core simply get small as the market gets large, even in simple markets?

Two approaches:

• Empirical: examine some simple markets
• Theoretical/computational: explore some artificial simple markets
The Thoracic surgery match is a simple match, with no match variations. It exactly fits the college admissions model; those theorems all apply.

**Descriptive statistics and original Thoracic Surgery match results**

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Applicant ROL’s</td>
<td>127</td>
<td>183</td>
<td>200</td>
<td>197</td>
<td>176</td>
</tr>
<tr>
<td>Active Programs</td>
<td>67</td>
<td>89</td>
<td>91</td>
<td>93</td>
<td>92</td>
</tr>
<tr>
<td>Program ROL’s</td>
<td>62</td>
<td>86</td>
<td>90</td>
<td>93</td>
<td>92</td>
</tr>
<tr>
<td>Total Quota</td>
<td>93</td>
<td>132</td>
<td>141</td>
<td>146</td>
<td>143</td>
</tr>
<tr>
<td>Positions Filled</td>
<td>79</td>
<td>123</td>
<td>136</td>
<td>140</td>
<td>132</td>
</tr>
</tbody>
</table>

**Difference in Thoracic Surgery results when algorithm changed from program proposing to applicant proposing:**

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>none</td>
<td>2 applicants improve 2 programs do worse</td>
<td>2 applicants improve 2 programs do worse</td>
<td>none</td>
<td>none</td>
<td></td>
</tr>
</tbody>
</table>
Theoretical computation on a simple model

Simple model: $n$ firms, $n$ workers, (no couples) uncorrelated preferences, each worker applies to $k$ firms.

$C(n) =$ number of workers matched differently at $\mu_F$ and $\mu_W$
Large core with $k=n$: $C(n)/n$ is the proportion of workers who receive different matches at different stable matchings, in a simple market (no couples) with $n$ workers and $n$ firms (each of which employs one worker) when preferences are uncorrelated and each preference list consists of all $n$ agents on the other side of the market. (from Roth and Peranson, 1999) Note that as the market grows large in this way, so does the set of stable matchings, in the sense that for large markets, almost every worker is effected by the choice of stable matching.
Small core of large markets, with \( k \) fixed as \( n \) grows: \( C(n)/n \) is the proportion of workers who receive different matches at different stable matchings, in a simple market with \( n \) workers and \( n \) firms, when each worker applies to \( k \) firms, each firm ranks all workers who apply, and preferences are uncorrelated. (from Roth and Peranson, 1999). Note that for any fixed \( k \), the set of stable matchings grows small as \( n \) grows large.
The numerical results show us that $C(n)/n$ gets small as $n$ gets large when $k$ is fixed, (even) for uncorrelated preferences.

And of course, in these simulated markets, we see that the core gets small not because of strategic behavior—these are the true preferences.

This also implies that in large markets it is almost a dominant strategy for every agent to reveal his true preferences—only one in a thousand could profit by strategically mis-stating preferences (if they had full information about all preferences).
In my class in the Spring of 2003, I posed the following as an Open Problem: **Prove**: The core gets small (in some expected value sense) as the market gets large, if the number of interviews doesn’t also get large.

Two of the students in that class did so... Nicole Immorlica and Mohammad Mahdian, “Marriage, Honesty and Stability,” Immorlica, SODA 2005, 53-62.

**Theorem (Immorlica and Mahdian, 2003:)**
Consider a marriage model with n men and n women, in which each woman has an *arbitrary* complete preference list, and each man has a random list of at most k women as his preference list (chosen uniformly and independently).
Then, the expected number of women who have more than one stable husband is bounded by a constant that only depends on k (and not on n). (So, as n gets large, the proportion of such women goes to zero...)
More recently, two former students have substantially generalized the result, to state that, as the market gets large, the opportunity for firms to manipulate either their preferences or their capacities gets small, in the sense that the proportion of firms with such an opportunity will go to zero.

**Theorem** (Kojima and Pathak, AER 2009): In the limit, as \( n \) goes to infinity in a regular sequence of random (many-to-one) markets, the proportion of employers who might profit from (any combination of) preference or capacity manipulation goes to zero in the worker proposing deferred acceptance algorithm.
Still (mostly) open empirical puzzle

• Why do these algorithms virtually always find stable matchings, even though couples are present (and so the set of stable matchings could be empty)?
Stylized facts

1. Applicants who participate as couples constitute a small fraction of all participating applicants.

2. The length of single applicants' rank order lists is small relative to the number of possible programs.

3. Applicants who participate as couples rank more programs than single applicants. However, the number of distinct programs ranked by a couple member is small relative to the number of possible programs.

4. The most popular programs are ranked as a top choice by a small number of applicants.

5. A pair of hospital programs ranked by doctors who participate as a couple tend to be in the same region.

6. Even though there are more applicants than positions, many programs still have unfilled positions at the end of the centralized match.

7. A stable matching exists in all nine years in the market for clinical psychologists.
Two initial approaches

• Kojima, Fuhito, Parag A. Pathak, and Alvin E. Roth, "Matching with Couples: Stability and Incentives in Large Markets"

• Stability in large matching markets with complementarities (previously called Matching with Couples - Revisited) Itai Ashlagi, Mark Braverman and Avinatan Hassidim (Extended abstract appears in EC 11.)
Random markets

• A random market is a tuple $\Gamma=\langle H,S,C, \succeq, \{\langle H^n, k,P,Q,\rho \rangle \rangle$, where $k$ is a positive integer (max length of ROL’s), $P=(p_{\{h\}})_{h \in H}$ and $Q=(q_{\{h\}})_{h \in H}$ are probability distributions on $H$, and $\rho$ is a function which maps two preferences over $H$ to a preference list for couples.

• Hospitals’ preference orderings are essentially arbitrary, and take account of their capacities, and couples preferences are formed from their individual preferences (drawn from probability distribution $Q$, different than $P$ for singles), via an essentially arbitrary function $\rho$. 
Random large markets

• A sequence of random markets is \((\Gamma^1, \Gamma^2, ...)\), where 
  \(\Gamma^n=(H^n, S^n, C^n, \geq_{\{H^n\}}, k^n, P^n, Q^n, \rho^n)\) is a random market in which 
  \(|H^n|=n\) is the number of hospitals.

Definition: A sequence of random markets \((\Gamma^1, \Gamma^2, ...)\) is **regular** if there exist \(\lambda>0, a \in [0,(1/2)), b>0, r\geq1,\) and positive integers \(k\) and \(\kappa\) such that for all \(n,\)

1. \(k^n=k\), \hspace{1cm} \text{(constant max ROL length, doesn’t grow with } n\)
2. \(|S^n| \leq \lambda n, \ |C^n| \leq bn^a\), \hspace{1cm} \text{(singles grow no more than proportionally to positions—e.g. } \lambda>1,\) \hspace{1cm} \text{and couples grow slower than root } n\)
3. \(\kappa_h \leq \kappa\) for all hospitals \(h\) in \(H^n\) \hspace{1cm} \text{(hospital capacity is bounded)}
4. \((p_h/p_{h'})\in [(1/r), r] \) \hspace{1cm} \text{and} \hspace{1cm} \((q_h/q_{h'})\in [(1/r), r]\) \hspace{1cm} \text{for all hospitals } h, h' \hspace{1cm} \text{in } H^n. \hspace{1cm} \text{(The popularity of hospitals as measured by the prob of being acceptable to docs does not vary too much as the market grows, i.e. no hospital is everyone’s favorite (in after-interview preferences))}
A (really simple) sequential couples algorithm (like left side of RP flow chart)

1. run a deferred acceptance algorithm for a sub-market composed of all hospitals and single doctors, but without couples,
2. one by one, place couples by allowing each couple to apply to pairs of hospitals in order of their preferences (possibly displacing some doctors from their tentative matches), and
3. one by one, place singles who were displaced by couples by allowing each of them to apply to a hospital in order of her preferences.

We say that the sequential couples algorithm succeeds if there is no instance in the algorithm in which an application is made to a hospital where an application has previously been made by a member (or both members) of a couple except for the couple who is currently applying. Otherwise, we declare a failure and terminate the algorithm.

Lemma: If the sequential couples algorithm succeeds, then the resulting matching is stable.
Stable matchings exist, in the limit

• Theorem: Suppose that \((\Gamma^1, \Gamma^2, \ldots)\) is a regular sequence of random markets. Then the probability that there exists a stable matching in the market induced by \(\Gamma^n\) converges to one as the number of hospitals \(n\) approaches infinity.
Key element of proof

• if the market is large, then it is a high probability event that there are a large number of hospitals with vacant positions at the end of Step 2 (even though there could be more applicants than positions) (i.e. the Scramble will remain important in large markets.)

• So chains of proposals beginning when a couple displaces a single doc are much more likely to terminate in an empty position than to lead to a proposal to a hospital holding the application of a couple member.
Formal statement

Proposition: many hospitals have vacancies in large markets

There exists a constant $\beta > 0$ such that

(1) the probability that, in a sub-market without couples, the doctor-proposing deferred acceptance algorithm produces a matching in which at least $\beta n$ hospitals have at least one vacant position converges to one as $n$ approaches infinity, and

(2) the probability that the sequential couples algorithm produces a stable matching and at least $\beta n$ hospitals have at least one vacant position in the resulting matching converges to one as $n$ approaches infinity.
Corollary

- Corollary: Suppose that \((\Gamma^1, \Gamma^2, \ldots)\) is a regular sequence of random markets. Then the probability that the Roth-Peranson algorithm produces a stable matching in the market induced by \(\Gamma^n\) converges to one as the number of hospitals \(n\) approaches infinity.
Stable Clearinghouses (those now using the Roth Peranson Algorithm)

NRMP / SMS:
Medical Residencies in the U.S. (NRMP) (1952)
Abdominal Transplant Surgery (2005)
Colon & Rectal Surgery (1984)
Combined Musculoskeletal Matching Program (CMMP)
• Hand Surgery (1990)
Medical Specialties Matching Program (MSMP)
• Cardiovascular Disease (1986)
• **Gastroenterology (1986-1999; rejoined in 2006)**
  • Hematology (2006)
  • Hematology/Oncology (2006)
  • Infectious Disease (1986-1990; rejoined in 1994)
  • Oncology (2006)
  • Pulmonary and Critical Medicine (1986)
  • Rheumatology (2005)
Obstetrics/Gynecology
• Reproductive Endocrinology (1991)
• Gynecologic Oncology (1993)
• Maternal-Fetal Medicine (1994)
• Female Pelvic Medicine & Reconstructive Surgery (2001)
Pediatric Cardiology (1999)
Pediatric Critical Care Medicine (2000)
Pediatric Emergency Medicine (1994)
Pediatric Hematology/Oncology (2001)
Pediatric Rheumatology (2004)
Pediatric Surgery (1992)

Primary Care Sports Medicine (1994)
Radiology
• Interventional Radiology (2002)
• Neuroradiology (2001)
• Pediatric Radiology (2003)
Surgical Critical Care (2004)
Thoracic Surgery (1988)

Postdoctoral Dental Residencies in the United States
• Oral and Maxillofacial Surgery (1985)
• General Practice Residency (1986)
• Advanced Education in General Dentistry (1986)
• Pediatric Dentistry (1989)
• Orthodontics (1996)
Psychology Internships in the U.S. and CA (1999)
Neuropsychology Residencies in the U.S. & CA (2001)
Osteopathic Internships in the U.S. (before 1995)
Pharmacy Practice Residencies in the U.S. (1994)
Articling Positions with Law Firms in Alberta, CA(1993)
Medical Residencies in CA (CaRMS) (before 1970)

****************
British (medical) house officer positions
• Edinburgh (1969)
• Cardiff (197x)

New York City High Schools (2003)
Boston Public Schools (2006)
Self-blocking couples: A different kind of (modeling) problem that we spent a little time thinking about is raised by the following partial example:

Let $C = \{a_1, a_2\}$ have preferences: $(H_1, H_2), (H_2, H_3)$

Suppose the relevant part of the hospital preferences are
- H1: $a_1, \ldots$
- H2: $a_1, a_2$
- H3: $a_2, \ldots$

Consider a (partial) allocation that has $C = \{a_1, a_2\}$ matched to H2, H3. That is, C gets its second choice, $[(a_1, H_2), (a_2, H_3)]$

Note that C would prefer the matching at which it was matched to H1, H2, i.e. $[(a_1, H_1), (a_2, H_2)]$

But $\{C, H_1, H_2\}$ don’t block the original matching because H2 prefers $a_1$ to $a_2$.

But maybe we should think of a modified definition in which $\{C, H_1, H_2\}$ does block the original matching, since C can withdraw $a_2$ from H2... As an administrative matter, C wouldn’t like to hear that they had gotten their second choice only because they had listed it as acceptable...
Scramble

• if the market is large, then it is a high probability event that there are a large number of hospitals with vacant positions at the end of a centralized match with short preference lists (even though there could be more applicants than positions) (i.e. the Scramble will remain important in large markets.)
Unmatched Applicants, Unfilled Positions
2001 - 2010

These charts do not fully depict the competitive nature of the Scramble. Many graduates of international medical schools (IMGs) do not submit rank order lists of programs; rather, they register for the Match to obtain the List of Unfilled Programs released during Match Week. The NRMP calculates that in 2010 nearly 13,000 applicants competed for only 1,060 first-year positions. The numbers are even more striking when one considers that more than 600 of the first-year positions were in preliminary programs, which many applicants view as undesirable because they do not lead to specialty training.
THE “SCRAMBLE” FOR UNFILLED POSITIONS

The “Scramble” officially begins on Tuesday at noon eastern time, when the NRMP posts the List of Unfilled Programs to its website, and continues until noon on Thursday, when U.S. medical schools hold their Match Day ceremonies. Although the unfilled positions remain posted until May 1 in the NRMP Registration, Ranking, and Results (R3) System, few are available after the first 48 hours:

Unfilled Positions: First Two Days

Data from the Electronic Residency Application Service (ERAS) corroborate the NRMP figures. Between noon and 5:00 p.m. on Tuesday, almost 80,000 MyERAS logins occurred, and more than 8,700 applicants sent just over 205,000 applications. Below are the numbers of applications submitted for the top seven specialties:

<table>
<thead>
<tr>
<th>Specialty</th>
<th>Mean Number of Applications Per Unfilled Program</th>
</tr>
</thead>
<tbody>
<tr>
<td>Anesthesiology</td>
<td>233</td>
</tr>
<tr>
<td>Emergency Medicine</td>
<td>382</td>
</tr>
<tr>
<td>Family Medicine</td>
<td>439</td>
</tr>
<tr>
<td>Internal Medicine</td>
<td>429</td>
</tr>
<tr>
<td>Pediatrics</td>
<td>221</td>
</tr>
<tr>
<td>Psychiatry</td>
<td>294</td>
</tr>
<tr>
<td>General Surgery</td>
<td>148</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Specialty</th>
<th>Applications Transmitted</th>
</tr>
</thead>
<tbody>
<tr>
<td>Anesthesiology</td>
<td>4,188</td>
</tr>
<tr>
<td>Emergency Medicine</td>
<td>1,910</td>
</tr>
<tr>
<td>Family Medicine</td>
<td>39,097</td>
</tr>
<tr>
<td>Internal Medicine</td>
<td>27,893</td>
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<td>5,077</td>
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<td>Psychiatry</td>
<td>3,816</td>
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<td>General Surgery</td>
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SCRAMBLE WORK GROUP

The Scramble Work Group established by the NRMP and AAMC identified several problems:

- No organization has stewardship of the Scramble.
- The Scramble lacks the trust, transparency, and integrity inherent in the Match.
- Applicants must make career decisions too quickly, often in minutes.
- There is no separation between application, interview, and appointment.
- There is no consistent or orderly process for applying to programs.
- No rules govern applicant and program behavior.

The Work Group agreed the NRMP is the logical organization to assume stewardship of the Scramble because it releases and manages the dynamic List of Unfilled Programs and controls which applicants have access to it. The Work Group also believed the NRMP could establish the same level of trust and integrity in the Scramble that exists in the Match. Accordingly, the Work Group recommended and the NRMP Board of Directors affirmed the following principles for change, which will serve as the basis for a Match Week Supplemental Offer and Acceptance Program (SOAP):

- Unmatched applicant and unfilled program information will be released simultaneously.
- There will be a “time out” period during which unmatched applicants can send applications but programs cannot make offers.
- Applicants and programs will be required to send and receive applications only through ERAS.
- NRMP-participating programs that fill positions during Match Week must do so only through the SOAP.
- New functionality will be added to the R3 System to allow programs to offer unfilled positions on the basis of preference lists submitted by the programs.
- Applicants must accept or reject their offer(s) within a specific timeframe; offers not accepted or rejected will expire.
- The R3 System will establish an electronic “handshake” when an applicant accepts a position.
- Positions will be deleted from the dynamic List of Unfilled Programs once an offer has been accepted.
- A program’s unfilled positions will be offered to applicants in order of preference until all positions are filled or the preference list has been exhausted; programs will be able to add applicants to the bottom of their preference lists throughout Match Week.
- The NRMP Match Participation Agreement will be expanded to include Match Week and SOAP, and sanctions will be imposed for improper behavior.
Comment on the NRMP’s “Supplemental Offer and Acceptance Program” Proposed to Replace the Post-Match Scramble

Peter A. Coles, Clayton R. Featherstone, John William Hatfield, Fuhito Kojima, Scott Duke Kominers, Muriel Niederle, Parag A. Pathak, and Alvin E. Roth

Executive Summary

Historic precedent and economic principles suggest that the Supplemental Offer and Acceptance Program (SOAP) proposed for the NRMP Scramble will lead to unsatisfactory outcomes by forcing participants to make unnecessarily difficult decisions and giving them strong incentives to break the rules laid out in the SOAP proposal. We suggest, as an alternative Scramble mechanism, that the NRMP run a “Second Match” for the Scramble participants using rules similar to those of the Main Match.

The SOAP Proposal

The SOAP proposal calls for up to eight rounds in which programs may make take-it-or-leave-it offers to applicants. During each round, programs send offers to applicants at the beginning of the round; no further offers may be made until the round is completed. At the end of each round, all current offers expire; hence applicants must decide whether to accept an offer without knowledge of which offers may be received in later rounds. In short, SOAP institutionalizes so-called exploding offers, that is, offers which are only valid for a fixed period, after which they “explode.”
Additional Strategic Concerns

Because of this urgency of matching in the first SOAP period, we anticipate strategic and potentially harmful applicant-program communications during the “time-out period.” We expect programs to solicit information about whether applicants are likely to accept offers, and may furthermore attempt to extract promises that a first round offer will be accepted. In turn, applicants will try to persuade programs that they will indeed accept first round offers. Both theoretical (Coles et al., 2010) and historic (Roth and Xing, 1997) evidence predict this behavior.

Misrepresentation by both programs and applicants may lead programs to make poor offer choices. For instance, programs may make offers to applicants who have indicated interest (possibly in response to inquiries from the program) but end up selecting another offer. Once again, poor strategic decisions by programs may lead to bad outcomes.

Out of fear that applicants will not accept first-round offers as promised, programs will have an incentive to coerce applicants into accepting early offers during the time-out period. One way to do this would be to make verbal contracts, with tacit or explicit threats that applicants not accept other offers during SOAP. Programs may also be able to work “within the system” to effectively make coercive offers during the time-out period. For example, they might instruct an applicant to withdraw from SOAP, verify that she has indeed done so, and then offer the applicant an “unfilled” position in the aftermarket.
Conclusion

In our opinion, the proposed SOAP mechanism will lead to coercive and strategic behavior, which will inhibit satisfactory outcomes. By contrast, we believe that a Second Match conducted among the Scramble participants will ensure an orderly and efficient matching of Scramble participants.
Monday, June 4, 2012

**First year of the new medical residency scramble, SOAP**

I've written before about the new Supplemental Offer and Acceptance Program (SOAP), and the National Resident Matching Program has now released a report on its first year of operation.

There were 1,100 unfilled first year positions at the end of the main match, and 815 unmatched seniors graduating from U.S. medical schools (and many more unmatched applicants when foreign medical schools are included). Most of the unmatched positions were in family medicine and in "preliminary" rotations in surgery and internal medicine.

After the first day of the SOAP exploding offer process (i.e. after two rounds of exploding offers), only 267 positions remained, and 98 of these remained unfilled. So, most of the action happened the first day.

Medical schools complained that students were asked to "commit" to programs prior to receiving an offer, and thought that rounds should be longer. Residency programs thought rounds should be shorter.

In line with the criticisms of the design offered earlier (see here), I anticipate that next year more students will be asked to "commit" before receiving an offer (even though it's against the rules), and that even more of the action will be concentrated in the first day and the first round, with more of the market shifting out of the formal scramble, either officially or de facto, through the offline "commitment" process....

As I was quoted saying last year (see here), "If it's really, really tempting for people on both sides to break the rules," says Roth, "often the rules get broken."