

Matching Mechanisms for Refugee Resettlement

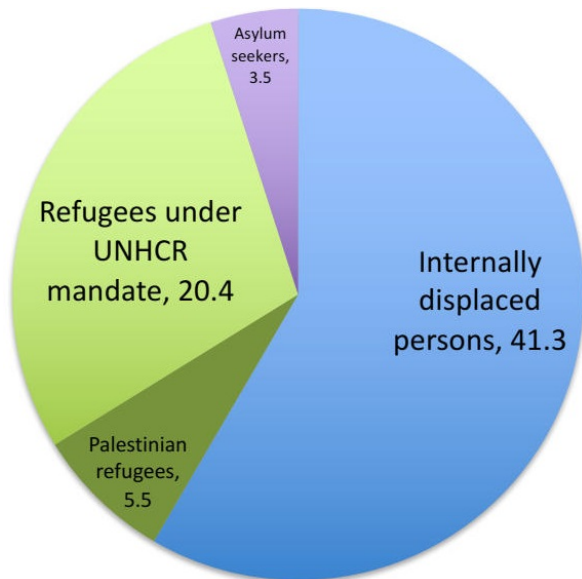
David Delacrétaz
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Scott Kominers
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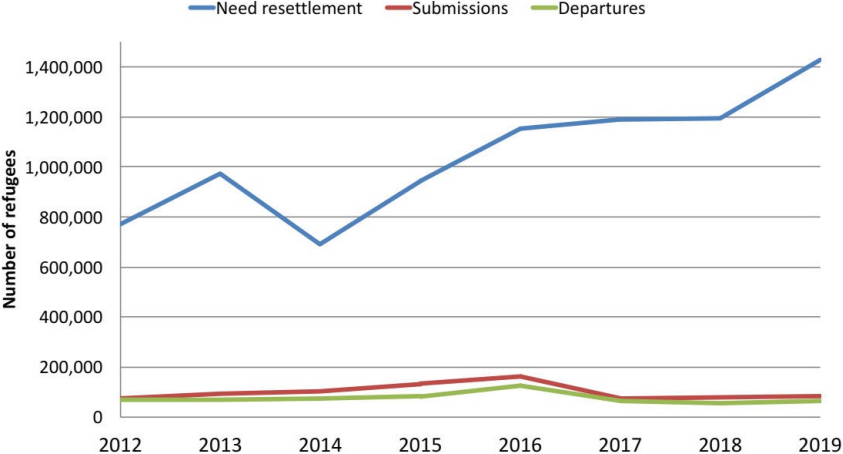
Alex Teytelboym
Oxford

Guest Lecture
AS.180.244 Market Design
AS.180.645 Topics in Economic Theory
Johns Hopkins University
May 2020

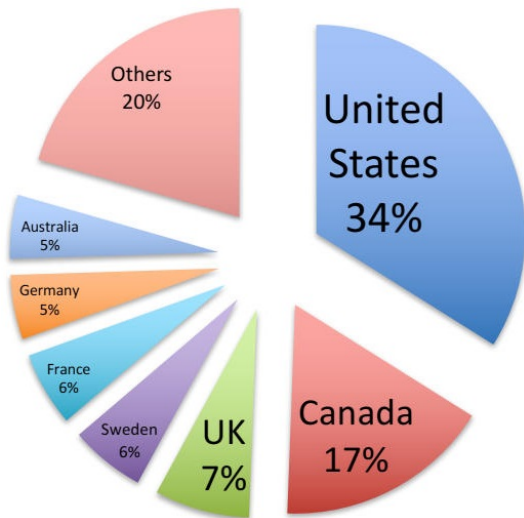
70M Refugees around the World



Resettlement Needs, Submissions, and Departures



Resettlement Destinations (2018)



Resettlement Process - USA

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- One of them is the Hebrew Immigrant Aid Society (HIAS)

The Matching Problem

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Figure: HIAS' network of localities.

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How to best match refugees and localities?

Why the matching matters

Empirical evidence that the initial match matters in for a refugee's long-term economic outcomes: employment, welfare, education

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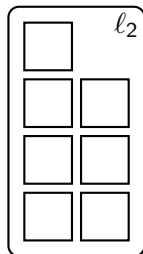
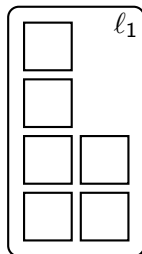
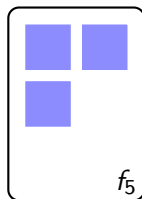
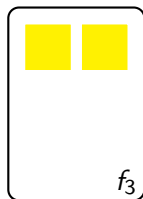
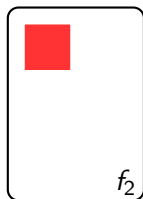
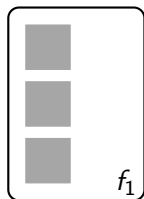
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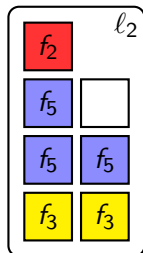
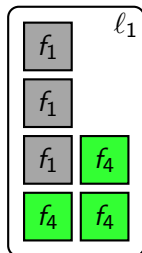
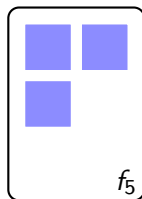
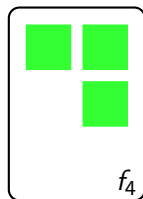
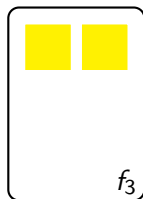
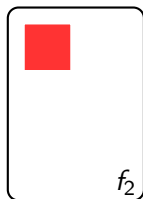
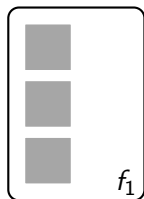
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Matching Market with Multidimensional Constraints

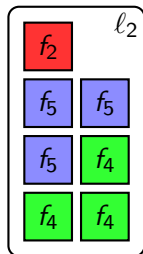
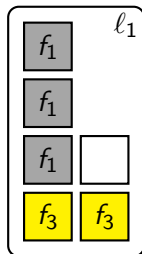
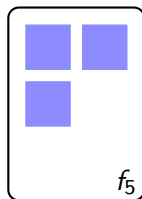
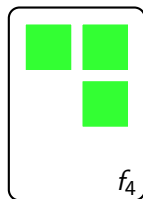
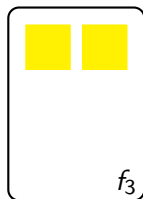
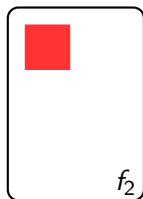
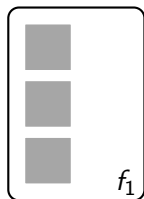
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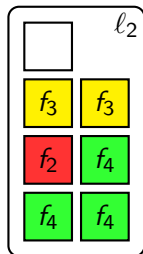
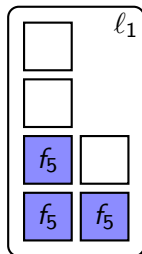
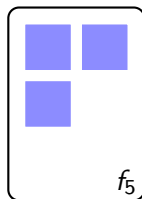
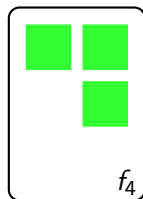
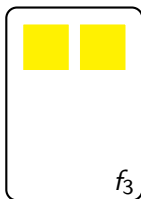
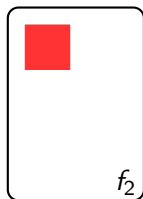
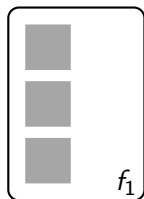
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We consider the static problem

- We take capacities as given and treat them as hard constraints
- Dynamic capacity management would constitute a valuable extension

Matching on Observables

1 May 2018: For the first time, a US resettlement agency used a software to algorithmically match refugees

- Trapp, Teytelboym, Martinello, Andersson, and Ahani (2018)

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For each family-locality pair, estimate a probability of employment

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- Other resettlement agencies do the matching by hand

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“Many Somali refugees initially settled around the country subsequently migrated to Lewiston, Maine. Lewiston has a weak economy but an established Somali community. Consequently, efforts to resettle these refugees elsewhere in the U.S. were less effective than they could have been. Their preferences should have been taken into account from the start.”

— Mark Hetfield (CEO of HIAS) in Roth (2015), “Migrants aren’t widgets”, *Politico*

Model

Set of **families** F , set of **localities** L

- Families have strict and ordinal **preferences** over localities

$$\succ_f : \ell_1, \ell_2, \ell_3, \dots$$

- Each locality strictly ranks families in order of **priority**

$$\triangleright_\ell : f_1, f_2, f_3, \dots$$

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Set of **services** S (# of refugees, school places, medical needs, etc)

- Family f requires $\nu_s^f \in \mathbb{Z}_{\geq 0}$ units of service s
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- Satisfies all **multidimensional constraints**:

$$\sum_{f \in \mu(\ell)} \nu_s^f \leq \kappa_s^\ell \quad \text{for all } \ell \in L \text{ and } s \in S$$

Plan

Part I: Improve upon an endowment

- Start with a matching (e.g., matching on observables)
- Use refugee preferences to find Pareto improvements
- Mechanism: **Multidimensional Top Trading Cycles with Endowment**

Part II: Account for refugee preferences and locality priorities

- Priorities come from administrative rules and/or preferences
- Solution concept: Weak Envy-freeness
- Trade-off between efficiency and strategy-proofness

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We use a modified version of Top Trading Cycles

Top Trading Cycles in School Choice

Suppose we are in the *school choice* environment

- $|S| = 1$ and $\nu_s^f = 1$ for all $f \in F$

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- Every family points at its most preferred locality
- Every locality point at its highest-priority family
- Every family in a cycle is matched to the family at which it is pointing
- The capacity of a locality that receives a family is reduced by one unit

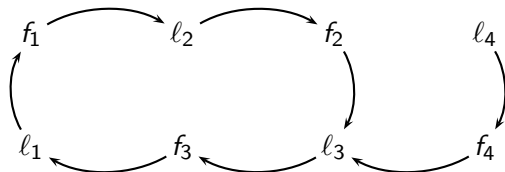
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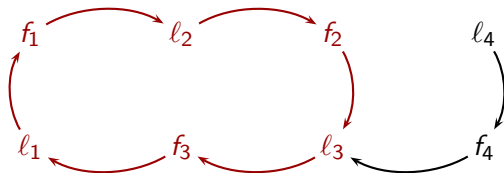
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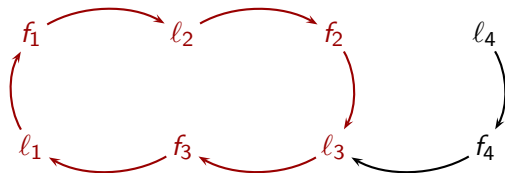
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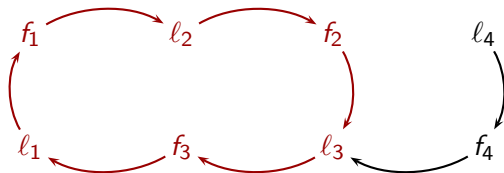
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- A priority can be interpreted as an endowment



Top Trading Cycles in Refugee Resettlement

TTC can easily be adapted to our environment

- **Multidimensional Top Trading Cycles (MTTC) mechanism**
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Improving upon an endowment is challenging

- Families of different sizes may not be able to swap with each other
- This problem does not occur in school choice

Pareto-improving Chains

Identifying Pareto-improvements is challenging

- Families with different needs may have to move simultaneously
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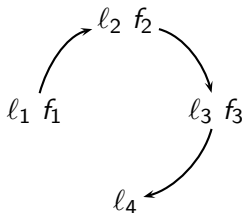
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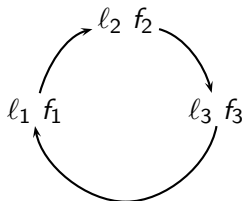
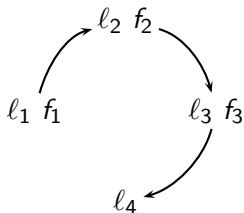
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 - moves to a locality where it can be accommodated (“open” chain) or
 - takes the place of the first one (“closed” chain)



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Definition

A matching is **chain-efficient** if it has no Pareto-improving chains

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Theorem

There does not exist any individually rational, chain-efficient, and strategy-proof mechanism.

Pareto Improvement

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When $|S| > 1$, there does not exist any strategy-proof mechanism that Pareto improves upon every chain-inefficient endowment.

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Theorem

When $|S| = 1$, there exists a strategy-proof mechanism that Pareto improves upon every chain-inefficient endowment.

Multidimensional Top-Trading Cycles with Endowment (MTTCE).

- Identifies and carries out chains to improve upon the endowment
- Individually rational and strategy-proof
- Pareto-improves upon any chain-inefficient endowment when $|S| = 1$

Plan

- Part I: Improve upon an endowment
 - Start with a matching (e.g., matching on observables)
 - Use refugee preferences to find Pareto improvements
 - Mechanism: Multidimensional Top Trading Cycles with Endowment

Part II: Account for refugee preferences and locality priorities

- Priorities come from administrative rules and/or preferences
- Solution concept: **Weak Envy-freeness**
- Trade-off between efficiency and strategy-proofness

Respecting Priorities

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Goodwill from localities matters

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How to account for both preferences and priorities?

- Stability is the natural solution concept

Example from McDermid and Manlove (2010)



Preferences

$f_1 : l_2, l_1, \emptyset$

$f_2 : l_1, l_2, \emptyset$

$f_3 : l_1, \emptyset$

Priorities

$l_1 : f_1, f_3, f_2$

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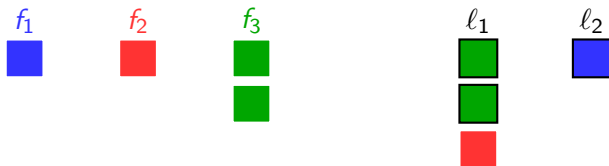
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Round 1		
f_1	\rightarrow	l_2
f_2	\rightarrow	l_1
f_3	\rightarrow	l_1

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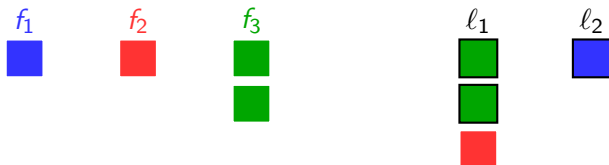
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Round 1				Round 2			
f_1	\rightarrow	l_2	✓	f_1	\rightarrow	l_2	
f_2	\rightarrow	l_1	✗	f_2	\rightarrow	l_2	
f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	

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f_2	\rightarrow	l_1	✗	f_2	\rightarrow	l_2	
f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	

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f_2	\rightarrow	l_1	✗	f_2	\rightarrow	l_2	✓
f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	✓

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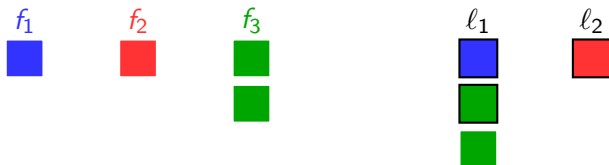
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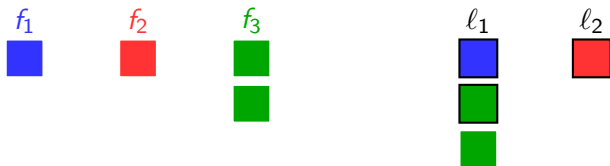
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f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	✗

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Round 1				Round 2				Round 3				Round 4			
f_1	\rightarrow	l_2	✓	f_1	\rightarrow	l_2	✗	f_1	\rightarrow	l_1	✓	f_1	\rightarrow	l_1	
f_2	\rightarrow	l_1	✗	f_2	\rightarrow	l_2	✓	f_2	\rightarrow	l_2	✓	f_2	\rightarrow	l_2	
f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	✗	f_3	\rightarrow	\emptyset	

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f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	✓	f_3	\rightarrow	l_1	✗	f_3	\rightarrow	\emptyset	✓

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If f_3 is matched to l_1

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If f_3 is matched to l_1

- (f_2, l_2) is a blocking pair unless f_2 is matched to l_2

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If f_3 is matched to l_1

- (f_2, l_2) is a blocking pair unless f_2 is matched to l_2
- But then f_1 remains unmatched and (f_1, l_1) is a blocking pair

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If f_3 is matched to l_1

- (f_2, l_2) is a blocking pair unless f_2 is matched to l_2
- But then f_1 remains unmatched and (f_1, l_1) is a blocking pair
- Therefore, f_3 is not matched to l_1 in any stable matching

Example from McDermid and Manlove (2010)



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- But then (f_3, l_1) is a blocking pair

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Stable matchings are not guaranteed to exist

Underusing capacities may be tolerable in refugee resettlement

- We propose a solution concept that respects priorities but may underuse some capacity

(Weak) Envy-freeness

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Given a matching μ , family g **envies** family f if g prefers f 's locality to its own and has a higher priority for it:

$$\mu(f) \succ_g \mu(g) \quad \text{and} \quad g \triangleright_{\mu(f)} f$$

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A matching is **envy-free** if no family envies another family

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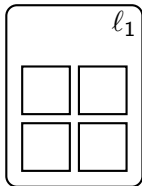
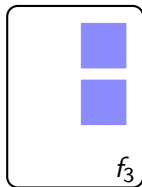
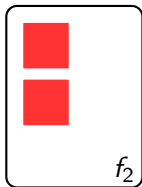
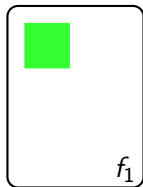
A matching is **envy-free** if no family envies another family

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Weak envy-freeness is a relaxation

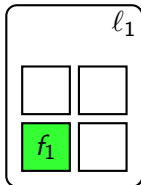
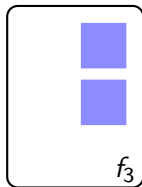
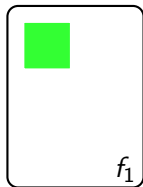
- A family f can be envied if it “fits” even when all families that envy f are matched to $\mu(f)$

Weak Envy-freeness



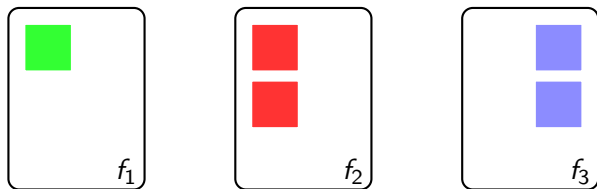
Priority: $f_1 \triangleright f_2 \triangleright f_3$

Weak Envy-freeness

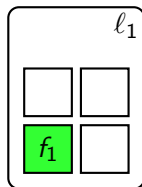


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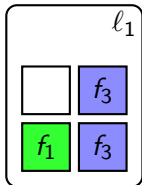
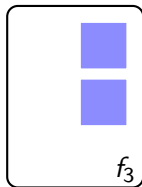
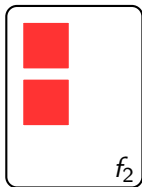
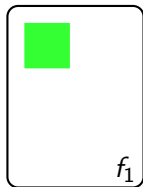


Envy-free and weakly envy-free



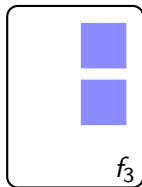
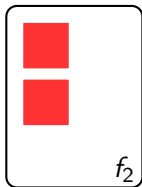
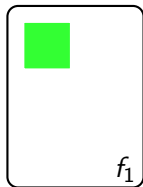
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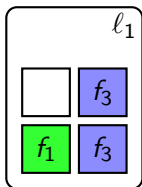


Priority: $f_1 \triangleright f_2 \triangleright f_3$

Weak Envy-freeness

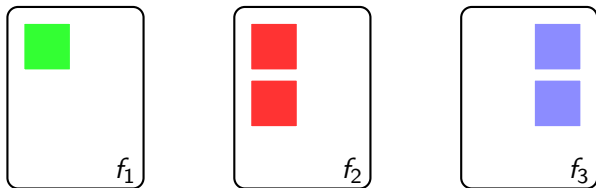


Not envy-free

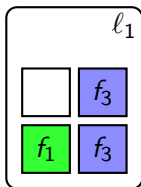


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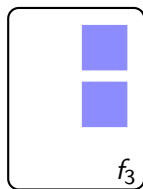
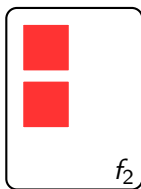
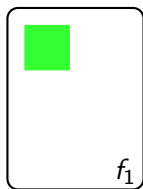


Not envy-free but weakly envy-free

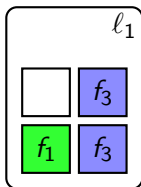


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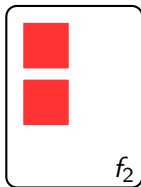
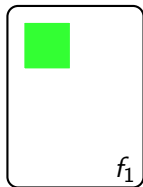


f_2 envies f_3 but f_3 fits even when f_2 is there

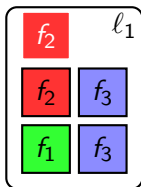


Priority: $f_1 \triangleright f_2 \triangleright f_3$

Weak Envy-freeness



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Theorem

There exists a unique family-optimal weakly envy-free matching and the CMDA algorithm finds it.

Strategy Proofness

The CMDA mechanism is not strategy-proof

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Solution: make the acceptance rule of localities harsher

- **Threshold Multidimensional Deferred Acceptance (TMDA)**
- Weakly envy-free and strategy-proof but not family-optimal

Conclusion

Refugee resettlement is a matching problem

- Optimally matching families and localities has long-term consequences
- Multidimensional constraints make it a complex matching problem
- Matching over observables has been done in practice

Solutions to account for preferences

- Using preferences has the potential to further improve the outcome
- Improvement over an endowment

Solutions to account for preferences and priorities

- Localities goodwill is important
- Solution concept: weak envy-freeness

Applications are just starting

- Applications will in turn inform theory