

# What can we do with MoNAn?

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An application to migration networks

# I FVFRHUI MF

### **Migration and mobility networks**

- Migration processes are characterized by interdependencies
  - Cumulative causation
  - Return migration
  - Migrant competition
  - Multilateral resistance to migration
- (Bilecen et al. 2018)



Yet, migration networks still understudied: a "missing link" in the literature



### Agency and structure in migration

Migration processes are inherently multilevel



**Micro:** individual characteristics, aspirations, motivations

#### Structure



**Macro:** origin and destination characteristics, international agreements

### Agency and structure in migration

Migration processes are inherently multilevel 

Agency



**Micro:** individual characteristics, aspirations, motivations

Leisure and lifestyle migration Retirement migration

#### Structure



**Macro:** origin and destination characteristics, international agreements

Organizational migration (Lucassen 2023) Soldiers, diplomats



### From structure to networks

- Structural factors particularly understudied (Findlay 2010)
  - Often due to data limitations: how do we measure institutions?
- Normally, we would need explicit measures
- Our proposed solution: to use networks to operationalize structure



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Degree student mobility









### **Background: international student mobility**

- - Long-term: **degree**
  - Short-term: credit, Erasmus+

International student mobility: people moving internationally for higher education

### **Background: international student mobility**

- - Long-term: **degree**
  - Short-term: credit, Erasmus+
- Evidence of inequalities in ISM:
  - Restaino, Primerano, & Vitale 2020, 2021)
  - Paulsen, & Pascarella 2010)

International student mobility: people moving internationally for higher education

Aggregate level: asymmetric patterns, core-periphery structures (Barnett et al. 2015; Brzenik & Skrbinjek 2020, Hou & Du 2020; Kondacki, Bedenlier, & Zawacki-Richter 2018;

Individual level: different participation rates by gender, with more males in degree mobility and more females in Erasmus+ (Böttcher et al. 2016; Salisbury,



### Research design: data

- 30 European countries
- All degree and Erasmus flows that occurred in 2019, disaggregated by gender
- **Degree:** 230,523 students
- **Erasmus:** 194,972 students
- **Gender breakdown:** Degree 56% women Erasmus 60% women



#### Total incoming students in 2019 (log)





### **Research design: methods**

- Degree and Erasmus mobility as weighted directed networks with countries as nodes and mobility flows as edges
- We apply a **MoNAn to assess** the impact of:
  - Individual-level factors (agency)
  - Endogenous and exogeneous factors (structure)











#### Networks of International Student Mobility in Europe, 2018-2019













### **Specifying effects**

# M&NAn Endogenous factors

#### Individual factors

• Gender

#### Exogenous factors

- Reciprocity
  - Concentration

- GDP per capita
- Population size
- Gender Inequality Index
- No. HEIs
- No. HEIs in Shanghai rankings
- No. ETPs
- Geographical distance
- Common language
- Migration flows
- Tourism flows
- Trade flows





### Full model

Covariate

GDP per capita similarity Population size similarity Gender Inequality Index simila English-taught programs (ETP Total HEIs similarity HEIs in Shanghai ranking simil Geographical distance Common language Migration flows Tourism flows Trade flows Geographical distance \* ge GDP per capita similarity \* get Reciprocity Concentration Reciprocity \* GDP per cap Concentration \* gender Country fixed effects Observations

Note: p < 0.05; p < 0.01; p < 0.01. Standard errors in parentheses.

	M4	M4
	Degree	Erasmus
	1.18***	1.55***
	(0.14)	(0.15)
	$0.17^{**}$	0.727***
	(0.074)	(0.087)
ritz	-0.733***	$0.184^{***}$
arity	(0.049)	(0.054)
Ps) similarity	0.0707	-0.172***
	(0.048)	(0.05)
	-0.231***	-0.759***
	(0.073)	(0.091)
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	-0.18***	-0.0473*
	(0.024)	(0.023)
	0.871***	-0.0594
	(0.023)	(0.046)
	0.0439***	-0.082***
	(0.0093)	(0.0091)
	0.175***	0.0316***
	(0.01)	(0.0083)
	-0.131***	0.0102
	(0.012)	(0.011)
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	(0.032)	(0.027)
ender	0.201	-0.067
	(0.17)	(0.14)
	0.0183***	
	(0.0023)	(0.0049)
		$0.0547^{***}$
	(0.003)	(0.0037)
pita similarity	-0.0159***	-0.029***
- 0	(0.0028)	(0.0059)
	<b>0.00487</b>	$0.0284^{\star\star\star}$
	(0.0075)	(0.0084)
	YES	YES
	20,000	$20,\!000$

Table 4: Results of the MoNAn (M4) models with complete specification

### Full model results I

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Covariate

GDP per capita similarity

Population size similarity

Gender Inequality Index similarit

English-taught programs (ETPs)

Total HEIs similarity

HEIs in Shanghai ranking similar

Geographical distance

Common language

Migration flows

Tourism flows

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### Full model results I

Table 4: Results of the MoNAn

Covariate

Trade flows

Geographical distance \* gend

GDP per capita similarity \* gende

Reciprocity

Concentration

Reciprocity \* GDP per capita

**Concentration** \* gender

Country fixed effects

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### **Results summary**

- beyond additional country-level and dyadic factors
- Main findings:

  - the degree mobility network
  - mobility and are more concentrated in their Erasmus moves

The MoNAn models allow us to calculate the effect of endogenous institutions,

Even when accounting for endogenous factors, difference in the quality of higher education between origin and destination affects ISM flows

Other "traditional" factors such as distance may not hold for Erasmus

Erasmus network is more reciprocal and less concentrated compared to

• There are gender effects where women travel shorter distances in degree



### Conclusion

- Across all model specifications, the Erasmus network exhibits stronger reciprocity and lower concentration compared to the degree mobility network
- The use of network effects to capture structural drivers appear promising
- The institutional fabric that underpins the Erasmus programme may be helpful in "rebalancing" ISM flows and countering asymmetric, core-periphery patterns
- This has implications for inequalities in ISM participation and broader phenomena such as the brain drain, even in highly developed countries





## I Dank your

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