

# The Global Race for Talent: Brain Drain, Knowledge Transfer and Growth by Marta Prato

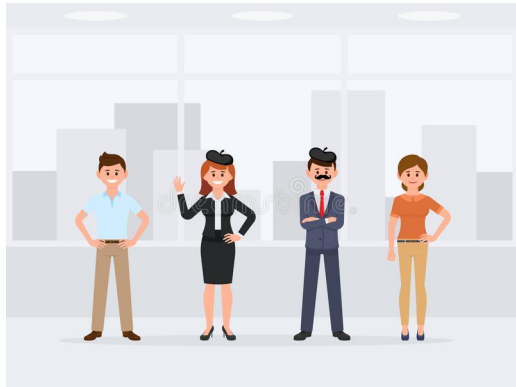
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THE UNIVERSITY  
*of* NORTH CAROLINA  
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# Human Capital Flight (Brain Drain) Isn't All Bad



It's also not entirely good. Equilibrium effects are complicated.

## Research Question

What happens when high-skill workers move?

# Theory and Empirics Can Connect Migration and Aggregate Growth

- Theoretical model, empirical estimation, and policy simulations to explore aggregate dynamics.
- Track effects at home- and host- country level:
  - **Allocation:** Where do talented workers flow?
  - **Innovation:** How do migrant and non-migrant inventors' ideas develop?
  - **Knowledge Transfer:** Do migrants facilitate cross-border knowledge diffusion?

# Paper Measures EU-US Migration Via Patent Data

- EU-US Corridor
- Track number of patents for migrant and non-migrant inventors and co-inventors
- Modify standard endogenous growth model to include migration decisions
- On balanced growth path, consider two policies:
  - Tax regimes, e.g., more profitable to invent in US
  - Migration caps, e.g., H1B program

# Plan

Data & Facts

Model

Inventors

Intermediates

Estimation

Counterfactual Analysis

Strengths of the Paper

Limitations and Extensions

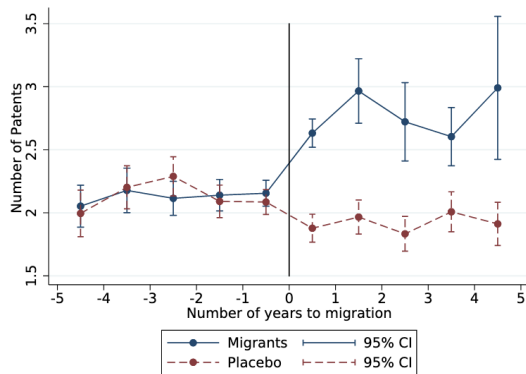
# Patent Data Is Both Rich and Imperfect

<b>Panel A: Number of Unique Observations</b>			
	Full Sample	EU Origin	US Origin
Unique Inventors	4,029,289	1,639,331	1,034,769
w/ more than 1 patent	1,293,431	593,328	344,938
Migrants	12,743	7,299	2,433
Return Migrants	2,371	1,350	475

- Patent dataset gives location, co-inventors, and citations
- Use last names to categorize inventor country of origin

# Yet Patent Counts Show Clear Trend Following Migration

Mean Number of Patents Before and After Migration Versus Placebo

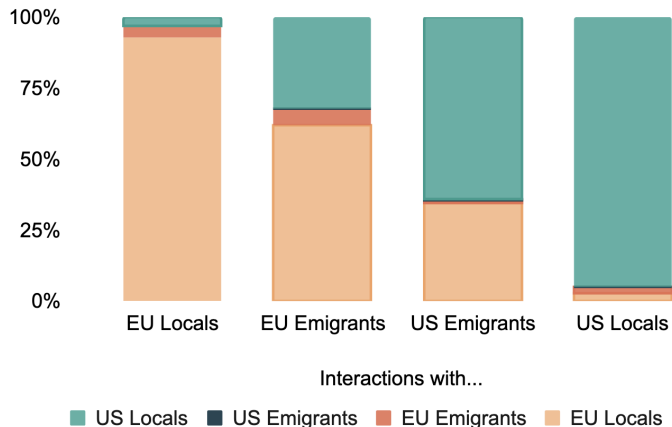


- Local co-inventors who never migrate patent more too!



# Migrants and Locals Have Different Interaction Networks

Figure 5: Interaction Networks



- Paper shows inventors access different networks post-migration.

## In Summary, Four Main Findings Motivate Model Design

1. EU has brain drain, US has brain gain
2. Migrants are more productive after migration
3. Migrant co-inventors are more productive after migration
4. Migrant interaction networks are more diverse after migration

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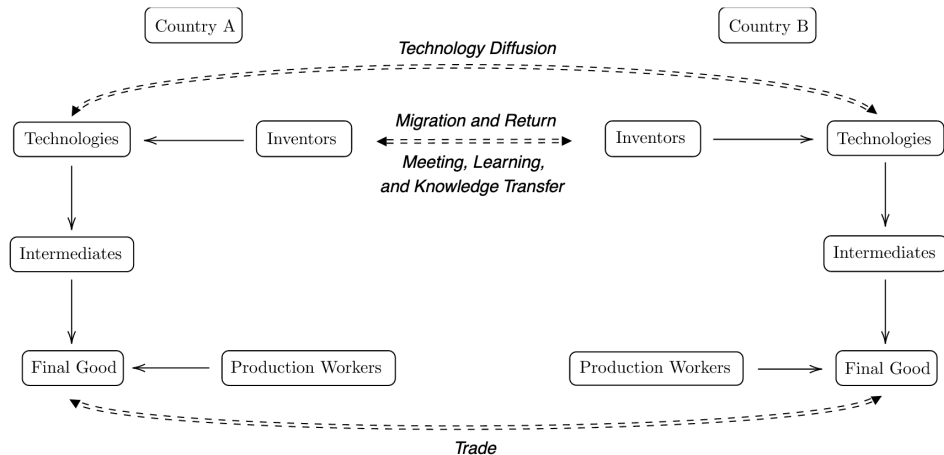
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# Use Endogenous Growth Model with a Few Tweaks

Figure 1: Summary of the Model



# Inventors: Inventors Are Heterogeneous on Two Dimensions

- Individual talent drawn from exogenous **country-specific** distribution:  
 $z \in \tilde{\mathcal{F}}_c, c \in \{EU, US\}$ .
  - By assumption,  $\tilde{\mathcal{F}}_{EU} = \tilde{\mathcal{F}}_{US}$ .
  - $z$  is an endogenous process.
- Individuals also draw **idiosyncratic, country-wide** productivity differential  $\varepsilon$  from an exogenous, country-specific distribution.
  - E.g., talented auto engineers might be better suited for Mercedes-Benz than, ahem, Ford.
  - $\varepsilon$  follows an exogenous, AR(1) process.
- Inventors produce ideas,  $q$ :

$$q(z, \varepsilon) = \begin{cases} z & \text{if local} \\ z + \varepsilon & \text{if migrant} \end{cases}$$

## Inventors: Inventor Talent Evolves Via Learning

- Every period, with probability  $\lambda$ , an inventor has a meeting with another inventor.
  - By construction, meetings **can only increase** an inventor's talent
  - Talent increase is proportional to partner's production bundle  $\tilde{q}$
- Inventors can meet with: locals, migrants from another country, or migrants from their country.
  - Introduce **meeting frictions**, i.e., locals more likely to meet locals

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- Inventors can meet with: locals, migrants from another country, or migrants from their country.
  - Introduce **meeting frictions**, i.e., locals more likely to meet locals
- Link back to 4 main findings:
  - Heterogeneity and learning  $\implies$  Brain Drain and increase in migrant production
  - Meeting frictions  $\implies$  networks more diverse following migration.

## Intermediates: Two Ways to Improve Product Quality

- Intermediate goods improve when:
  1. Monopolists purchase technology  $q$  from inventors
  2. Monopolists in **laggard** economy (i.e.,  $\varepsilon_c$  is lower) receive exogenous **technology spillover** from frontier economy
- Purchases: Monopolists matched in market to inventors  $\implies$  crowding out possible
- Spillover: spillover size is proportional to average quality difference between economies



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- Spillover: spillover size is proportional to average quality difference between economies
- Link back to 4 main findings:
  - Knowledge spillovers  $\implies$  co-migrants become more productive.

# Migrant Maximizes Over Value of Staying and Value of Migrating

Value of Staying:

$$\begin{aligned} V_{EU,EU}(z, \varepsilon, t) = & \pi_{EU}(z, t) \\ & + \beta \delta \int_{-\infty}^{\infty} \left( \lambda \sum_j \psi_{EU,EU,j,t} \int_1^{\infty} (W_{EU,EU}(z \tilde{q}^{\eta}, \varepsilon', t+1)) dF_{j,t}(\tilde{q}) \right. \\ & \left. + (1 - \lambda) W_{EU,EU}(z, \varepsilon', t+1) \right) dv_{\varepsilon'|\varepsilon} \end{aligned}$$

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Current period profits

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Continuation Value

# Migrant Maximizes Over Value of Staying and Value of Migrating

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Successful meeting

# Migrant Maximizes Over Value of Staying and Value of Migrating

$$\begin{aligned} V_{EU,EU}(z, \varepsilon, t) = & \pi_{EU}(z, t) \\ & + \beta \delta \int_{-\infty}^{\infty} (\lambda \sum_j \psi_{EU,EU,j,t} \int_1^{\infty} (W_{EU,EU}(z \tilde{q}^{\eta}, \varepsilon', t+1)) dF_{j,t}(\tilde{q})) \\ & + (1 - \lambda) W_{EU,EU}(z, \varepsilon', t+1)) dv_{\varepsilon'|\varepsilon} \end{aligned}$$

Choice of whether to migrate next period

# Migrant Maximizes Over Value of Staying and Value of Migrating

Value of **Migrating**:

$$\begin{aligned} V_{EU,US}(z, \varepsilon, t) = & \pi_{US}(z + \varepsilon, t) \\ & + \beta \delta \int_{-\infty}^{\infty} \left( \lambda \sum_j \psi_{EU,US,j,t} \int_1^{\infty} (W_{EU,US}(z \tilde{q}^{\eta}, \varepsilon', t + 1)) dF_{j,t}(\tilde{q}) \right. \\ & \left. + (1 - \lambda) W_{EU,US}(z, \varepsilon', t + 1) \right) dv_{\varepsilon'|\varepsilon} \end{aligned}$$

**Profits include epsilon**

# Establish a Link Between Individual Migration Decisions and Aggregate Growth

## Individuals

- Inventors maximize over value of staying, and value of going. Migration has a cost.
- When migrating, three things to consider:
  1. Higher TFP  $\implies$  higher inventor profits
  2.  $q$  (tech. bundle) is a function of  $\varepsilon$
  3. Change in meeting frictions as immigrant, potential to make higher quality matches

## Aggregates

- TFP is a function of quality of technology in a country.
- For laggard economy, knowledge spillover is proportional to TFP gap.
- Matches between intermediates and inventors can crowd out innovation by locals.



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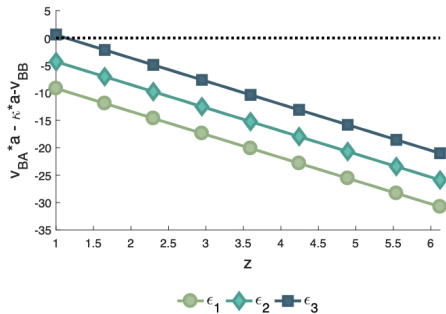
# The Model Matches the Data Very Well

Table 6: Moments

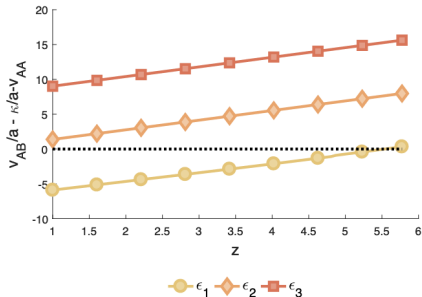
Moment	Data	Model
Share Migrants EU-US	6.00	6.83
Share Migrants US-EU (% domestic inventors)	0.40	0.39
Share Return Migrants (% migrants)	0.13	0.10
$\Delta$ productivity migrants EU-US (%)	0.28	0.32
$\Delta$ productivity co-inventors of migrants EU (%)	0.17	0.16
$\Delta$ productivity co-inventors of migrants US (%)	0.19	0.18
Growth rate (%)	1.50	1.39
TFP gap	0.90	0.90

# Value of Migration Much Higher for EU Locals

## Value of Migration for US Locals



## Value of Migration for EU Locals



► Data Compared to Event Study

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**Counterfactual Analysis**

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# Two Policy Scenarios to Replicate Real-World Policies

## **Tax Cuts**

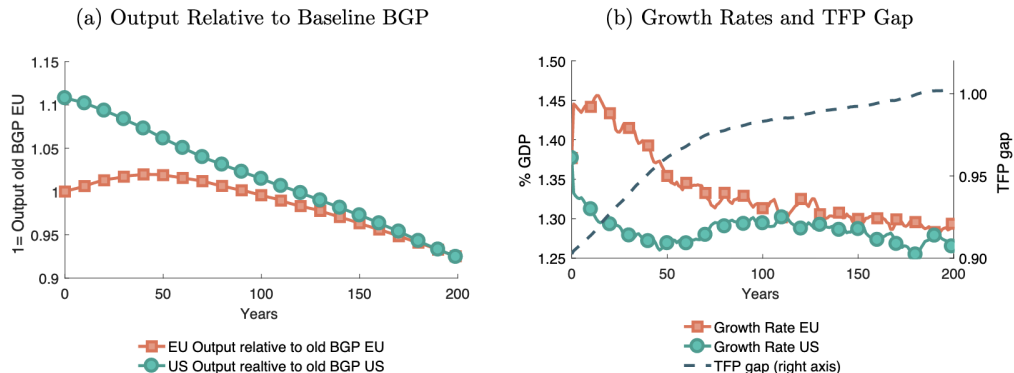
- Along BGP, EU has higher tax rates than US
- To stem brain drain, lower taxes for inventors to migrate and return home
- Eliminate tax gap between US and EU

## **Migration Caps**

- Along BGP, US has baseline migration cap of 1%
- If demand for migration exceeds cap, lottery to determine who gets a spot
- US selects most talented individuals and doubles migrant cap

# Tax Cuts to Reverse Brain Drain Equalize TFP at Cost of Total Output

Figure 13: Tax Cut for Foreigners and Return Migrants in the EU: Transitional Dynamics.

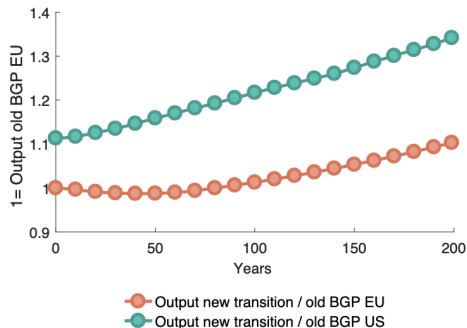


**Fixing brain drain is hard**

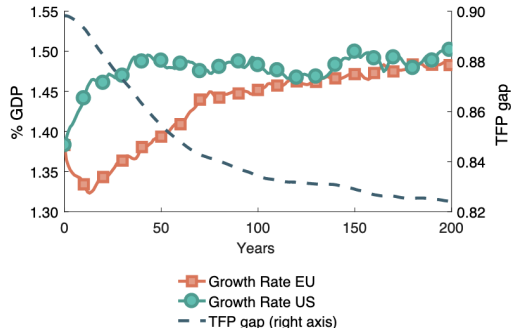
# Reduced Migration Cap and Talent Selection Spurs Global Growth

Figure 17: Counterfactual Increase of US Migration Threshold: Transitional Dynamics.

(a) Output Relative to Baseline BGP



(b) Growth Rates and TFP Gap



**Free markets rule!**

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**Strengths of the Paper**

Limitations and Extensions



# Overview of Strengths of the Paper

- Relevant question
- Model with many layers and elements
- Empirics motivate model well

## Strength: Relevance

The research question itself is interesting and important. Beyond that, this paper is also related to 6 strands of literature:

- Endogenous growth
- Human-capital-based growth
- Allocation of talent and relationship to growth
- Empirical study of knowledge diffusion
- Link between innovation, migration, and growth
- Effects of taxation on migration flows and innovation

## Strength: Model

The model in this paper contains many layers:

- Innovation-based endogenous growth
- Heterogeneous agent
- Two countries setting
- Network of knowledge diffusion
- etc

which makes the analysis of effects of migration on various issues possible.

# Strength: Model

## Existing models

- either study only micro-level migration decisions, taking macro environments as given.
- or study only macro effects of immigration, taking migration flows as given.

## This paper

- introduces endogenous migration in innovation-based growth model, allowing analysis of impact of policies.

## Strength: Empirics

- The stylized facts motivate very well for a model
- Creative way of identifying country of origin
- Used micro-data to pinned down meeting frictions

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## Limitation: Mobility data

- No direct information on migration from patents
- Namsor (software that analyzes ethnic origin of names, via algorithmic search of administrative databases) used to return a likely country of origin
  - Approach might not distinguish foreign-born researchers and later generations of foreign researchers
- Empirical results that rely on individual flows depend on accuracy of software

## Limitation: Patents

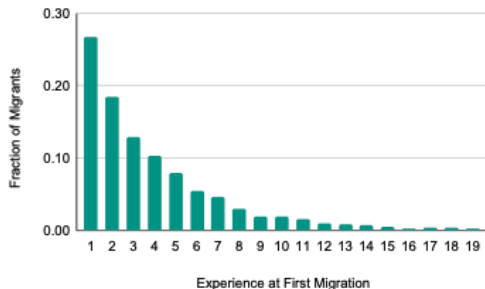
- Productivity or innovative output measured by number of patent applications
- Migrants can only be identified after at least one patent, but this excludes those who move before ever filing
- Also cannot track returns unless there are at least three patent applications
- Difficult to show any causal effects
  - Inventors move when they have a good draw of productivity
  - Not necessarily a causal effect of migration on innovative activity



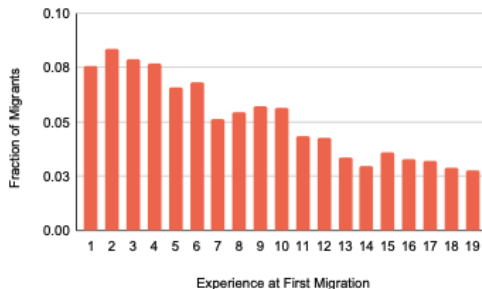
## Limitation: Experience at migration

Figure 9: Experience at First Migration: Data vs. Model

(a) Data



(b) Model

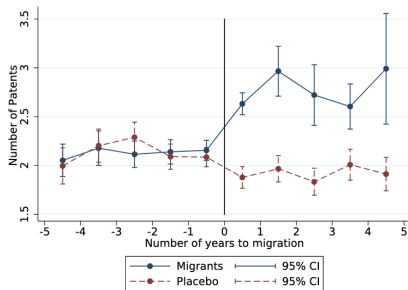


## Limitations: Geographical differences

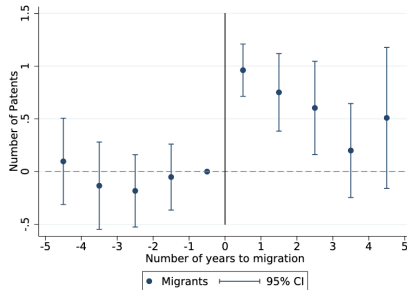
- Quantitative analysis starts from two countries with identical parameters and different policies
- There is room to include more country-specific heterogeneity; migration dynamics and knowledge spillovers could be very different
  - Country A is an emerging economy, country B is a developed economy
  - The countries have comparative advantages in different industries
- Would the results still be the same if we limited data to migrants moving between densely populated/urban areas?

# OLS Supports Change In Behavior

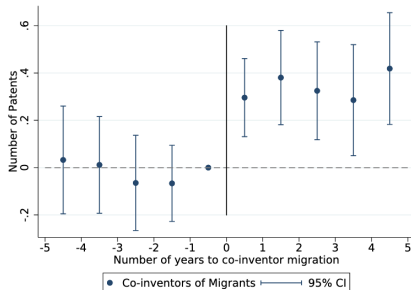
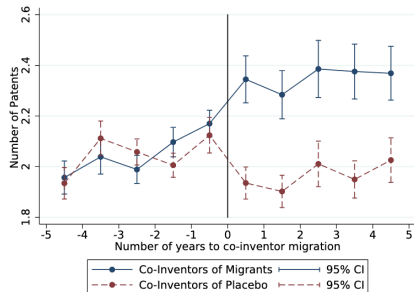
(a) Raw Means



(b) Coefficients  $\beta_{\tau}^{Mig}$  for migrants



# Co-Inventors Increase Patents After Migration Too



[Return](#)

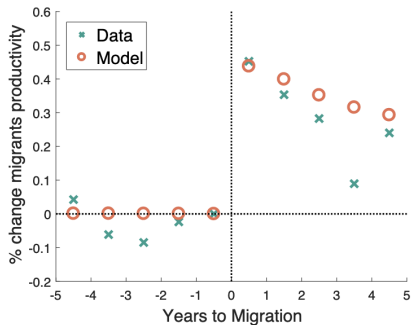
# Calibrated Moments

Table 5: Parameter Values

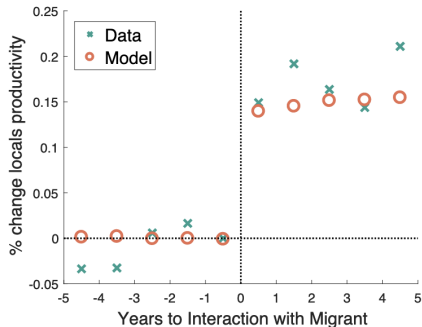
Parameter	Description	Value
— Panel A. External Calibration —		
$\beta$	Discount Rate	0.97
$\tau$	Interest Rate	0.03
$\delta$	Survival Rate	0.95
$\alpha$	Final Good Production	0.11
$\nu$	Inventor-Firm match rate	1.00
$\tau_A$	Tax Rate EU	0.40
$\tau_B$	Tax Rate US	0.30
$I_A$	Share R&D workers	0.01
— Panel B. Direct Match to Data —		
$\xi_{AB,AA}$	Meeting Frictions	1.31
$\xi_{AB,BB}$	Meeting Frictions	0.65
$\xi_{BB,AA}$	Meeting Frictions	0.06
$\xi_{BA,AA}$	Meeting Frictions	0.71
$\xi_{BA,AB}$	Meeting Frictions	0.32
$\xi_{BA,BB}$	Meeting Frictions	1.24
— Panel C. SMM Calibration —		
$\bar{\mu}$	Migration cap to US (Share of Inventors)	0.01
$\kappa$	Cost of Migration	0.10
$\lambda$	Meeting Intensity HH	0.10
$\eta$	Learning Technology	0.34
$\sigma$	Technology Absorption	0.02
$\theta_A$	Talent CDF H	15.00
$\rho_A$	Location Shock Persistence H	0.89
$\omega_A$	Location Shock SD H	0.20

# Simulated Data Compared to Event Study

(a) Migrants around Time of Migration



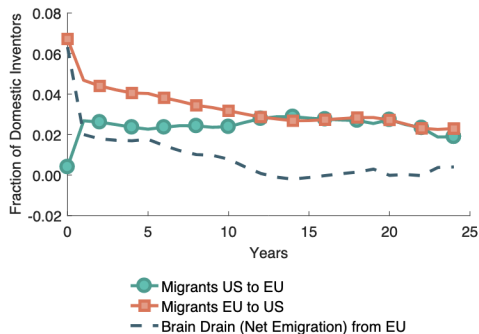
(b) Locals around Interaction with Migrant



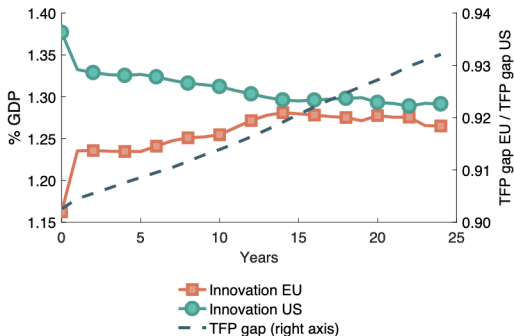
# Tax Cut Transition Dynamics Look Better for EU

Figure 12: Tax Cut for Foreigners and Return Migrants in the EU: Transitional Dynamics.

(a) Migrant Inventors by Nationality



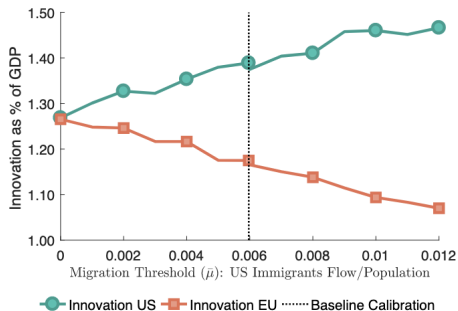
(b) Innovation and TFP gap



# Migration Limit Hurts US GDP Growth

Figure 14: Counterfactual Change to US Immigration Threshold ( $\bar{\mu}$ ): BGP Comparison

(a) Innovation and TFP gap



(b) Stock of Migrants

