The Impact of Immigration on the Internal mobility of Natives and Foreign-born Residents: Evidence from Italy

Presenter: Alessandra Faggian Gran Sasso Science Institute, L'Aquila – Italy

Co-authors: Roberto Basile (University of L'Aquila) - Luca De Benedictis (University of Macerata) - María Durban (Carlos III) - Román Mínguez (Castilla-La Mancha)

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Background and Motivation

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- European countries (and not only) are a major destination of international migrants (Münz, 2007) and simulations indicate that this will continue in the decades to come (Hanson and McIntosh, 2016; Docquier and Machado, 2017)
- The main worry about international migrants is always: how will they affect the native population? And, in particular, their effects on the domestic labour market

Are immigrants "stealing" the jobs of the natives and displacing them?



Dal Mediterraneo ai ristoranti di Milano, elogio degli "immigrati che rubano il lavoro agli italiani"

La storia di Daouda, ivoriano, che è stato assunto in uno dei migliori ristoranti di Milano grazie al progetto Sprar di Trezzano sul Naviglio. E altri commercianti ora si stanno interessando al progetto

di Lidia Baratta





22 ottobre 2016

Le verità scomode su migranti e lavoro

Una ricerca del Cer sfata un tabù della sinistra europea. Nei Paesi del Sud gli stranieri fanno concorrenza ai locali. Al Nord pesano sul y welfare

ENRICO PEDEMONTE

Analisi Gli stranieri coprono i buchi, non rubano i nostri posti di lavoro

Awenire

FRANCESCO SEGHEZZI E MICHELE TIRABOS

Gli immigrati rubano il lavoro agli italiani? E per questo motivo è necessario impedire o comunque limitare il loro arivo? È questa una delle modalità in cui oggi si usa rappresentare l'epocale fenomeno migratorio al quale stiamo assistendo ormai da alcuni anni, alimentando paure, rancori e sentimenti avversi.

A PAGINA 3



I migranti coprono i buchi non rubano posti di lavoro Occupazione e salari, la concorrenza è tra stranier.

ANALISI / COSA DICONO I DATI AL DI LÀ DELLA PERCEZIONE COMUNE

studi scientifici in materia confermano un quadro diverso da quello che viene presentato nel dibattito pubblico. In primo luogo i migranti si concentrano spesso e quasi unicamente in mestieri e settori professionali diversi e Michele Tiraboschi da quelli ambiti e occupati dai lavoratori del Paesi The political debate often revolves around this question, but the answer depends on our beliefs about the **"degree of subtitutability"** between migrants and natives in the labour market



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If natives and immigrants are **perfect substitutes** in the labour market, then the increase in labour supply leads to lower wages, which lead to native out-migration (**displacement of natives**)



Labor supply increases from S0 to S1 due to immigration. To the extent that S2 shifts in from domestic outmigration, there are small *ex post* wage and employment effects. (S3 could occur if there are added *cultural avoidance effects (Faggian et al. 2013)*, causing higher wages and less employment).

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If natives and immigrants are **complements** in the labour market, then the effect could actually be positive on wages and employment leading to internal in-migration rather than out-migration



Labor supply increases from S0 to S1 due to immigration. If strong complements, then D0 shifts to D1. The offsetting wage increase could in fact spur in-migration of native workers causing an additional increase in supply beyond S1. This causes a highly elastic population response.



Even when immigration is concentrated in few gateway areas, internal migration flows allows the effects to spread across the country ('bathtub model', Borjas, 2003 & 2005; Partridge and Rickman 2008, 2009)

In the USA:

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- Borjas (2003) and Partridge et al. (2008, 2009) find strong domestic resident response to international immigration.
- However, Card and DiNardo (2000) Card (2001) and Kritz and Gurak (2001) find that there is little connection between foreign immigration and native out-migration.

Card and DiNardo (2000): they even say that an increase in immigrants in **certain skill groups** lead to a *small increase* in the native population of the similar skill groups (**complementarity**)



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For the case of Italy – similarly to Card and DiNardo - Mocetti and Porello (2010) find that immigration is positively associated with infows of highly-educated natives (complementarity effect), and negatively associated with inflows of loweducated natives (displacement effect)

However, this study (and the previous ones) does not consider the *effect of foreign immigrants on internal movements of previous immigrant cohorts*, a phenomenon gaining momentum in recent years



Annual flows of foreign immigrants and of interprovincial movements of natives and foreign-born residents





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9 Aim & contribution

<u>Aim</u>: Estimating a gravity model of internal migration across Italian provinces to test the effect of new foreign immigrants on internal movements of **both natives** (with different educational levels) and of **foreign-born residents**

Contribution:

- 1. We look also at foreign-born residents (hp: different degrees of substitutability with immigrants)
- 2/We look at a finer spatial scale (NUTS3, rather than NUTS2)
- 3. We look at a longer time span 2002-2011
- 4. We propose a different empirical specification (negative binomial) to account for very high overdispersion in the data
- 5. We control for origin & destination fixed effects simultaneously by including smooth spatio-temporal trends

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Data

We use official data on internal mobility at the province (NUTS-3) level collected by ISTAT ("Indagine sui trasferimenti di residenza") for the period 2002-2011.

Four levels of schooling: 1 primary school 2 lower-secondary school 3 upper-secondary school 4 higher (i.e. Bachelor's degree, Master's degree, and PhD)

We select movers in the age class 15-64 years.

For the case of foreign-born residents and of new immigrants, there is no information on the educational attainment of movers.



Immigration rates from abroad of foreign workers

2003







Share of foreign immigrants over total population of the province



Internal mobility rates of foreign-born residents

North-South divide...

2003



But...in 2011 also out-migration from North-East (industrial districts)

2011





Modelling

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1. Baseline specification with explanatory variables in relative terms

$$\eta_{jkt} = \log(\mu_{jkt}) = \beta_1 \log\left(\frac{lmm_{kt}}{lmm_{jt}}\right) + \beta_2 \log(\phi_{jk}) + \beta_3 \log\left(\frac{y_{kt}}{y_{jt}}\right)$$

$$\mu_{jkt} = E\left(m_{jkt}^h\right) \quad m_{jkt}^h \sim l \quad gbin(\mu, \theta)$$
Number of migrants (with educational level h in the cas of Italian residents) moving at time t from province of origin j to province of destination k
$$M_{jkt} = \frac{1}{2} \log\left(\frac{mm_{kt}}{mm_{jt}}\right) + \beta_2 \log(\phi_{jk}) + \beta_3 \log\left(\frac{y_{kt}}{y_{jt}}\right)$$

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2. Specification including spatio-temporal trends

$$\eta_{jkt} = \log(\mu_{jkt}) = \beta_1 \log\left(\frac{lmm_{kt}}{lmm_{jt}}\right) + \beta_2 \log(\phi_{jk}) + \beta_3 \log\left(\frac{y_{kt}}{y_{jt}}\right)$$
$$f^j(x_{s1}, x_{s2}, x_{\tau}) + f^k(x_{s1}, x_{s2}, x_{\tau})$$
$$\mu_{jkt} = E\left(m_{jkt}^n\right) \quad m_{jkt}^n \sim Negbin(\mu, \theta)$$

<u>Problem</u>: In standard gravity models, the way to *correct for relevant omitted variables* (e.g. natural amenities, regional business cycles) is to include *origin and destination fixed effects*. However, in our model including origin-by-year nd destination-by-year fixed effects would have meant estimating 2,069 incidental parameters. *With count data models*, this approach is usually *problematic* since negbin withn so many parameters can hardly converge....

<u>Solution</u>: using smoothing functions that are function of space (latitude and longitude) and time. Alternative to dummies as long as unobserved heterogeneity is smoothly distributed over space...

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 Now, the assumption that unobserved spatial heterogeneity is smoothly distributed over space would be a problem in most countries, but Italy is "special", because of its famous North-South divide...

(f) 2003

(g) 2011

FIGURE 11 Disposable income per capita

TABLE 2

Results (1)

Gravity models of internal mobility in Italy. Testing the relative incidence of foreign immigrants.

Negative Binomial Estimates. Baseline specification (1)

	Italian citizens				Foreign-born	
	Primary	Lower sec.	Upper sec.	Higher	residents	
Parametric terms	Coefficients (standard errors)					
Intercept	-5.675***	-4.056***	-3.959***	-2.323***	-1.582***	
	(0.031)	(0.024)	(0.022)	(0.025)	(0.023)	
$\log(Imm_{kt}/Imm_{jt})$	-0.037***	0.001	0.064***	0.077***	-0.113***	
	(0.014)	(0.010)	(0.009)	(0.011)	(0.011)	
$\log(Inc_{kt}/Inc_{jt})$	-0.090**	0.067**	0.300***	0.661***	0.526***	
	(0.040)	(0.028)	(0.027)	(0.031)	(0.029)	
$u_{kt} - u_{jt}$	-0.020***	-0.019***	-0.015***	-0.012***	-0.038***	
-	(0.002)	(0.001)	(0.001)	(0.001)	(0.001)	
$\log(House_{kt}/House_{jt})$	0.054***	0.082***	0.082***	0.088***	0.071***	
	(0.015)	(0.011)	(0.010)	(0.012)	(0.011)	
$\log(Agr_{kt}/Agr_{jt})$	0.062***	0.034***	0.015***	-0.027***	0.027***	
-	(0.006)	(0.004)	(0.004)	(0.005)	(0.004)	
$\log(Cons_{kt}/Cons_{jt})$	0.024	-0.031*	-0.027	-0.044*	-0.025	
-	(0.027)	(0.019)	(0.018)	(0.021)	(0.020)	
$\log(Man_{kt}/Man_{jt})$	-0.047***	-0.026***	0.004	-0.010	0.021**	
· • •	(0.023)	(0.008)	(0.008)	(0.009)	(0.008)	
$\log(\phi_{jk})$	-0.574***	-0.619***	-0.638***	-0.665***	-1.100***	
v	(0.005)	(0.004)	(0.004)	(0.004)	(0.004)	

Displacement: lowest educated natives & other foreign-born

Complementarity: more educated natives

TABLE 3

Gravity models of internal mobility in Italy. Testing the relative incidence of foreign immigrants. Negative Binomial Estimates. Model specification (2) with spatio-temporal trends

Results (2)

	Italian citizens				Foreign-born	
	Primary	Lower sec.	Upper sec.	Higher	residents	
Parametric terms	Coefficients (standard errors)					
Intercept	-2.462***	-0.830**	-1.269***	-0.914***	-2.552***	
	(0.486)	(0.399)	(0.323)	(0.284)	(0.229)	
$\log(Imm_{kt}/Imm_{jt})$	-0.041**	-0.032**	0.006	-0.006	-0.130***	
	(0.019)	(0.014)	(0.013)	(0.016)	(0.014)	
$\log(Inc_{kt}/Inc_{jt})$	-0.013	-0.008	0.121***	0.325***	0.020	
-	(0.058)	(0.043)	(0.039)	(0.047)	(0.039)	
$u_{kt} - u_{jt}$	-0.012***	-0.011***	-0.010***	-0.011***	-0.023***	
-	(0.003)	(0.002)	(0.002)	(0.002)	(0.002)	
$\log(House_{kt}/House_{jt})$	-0.012	-0.004	-0.002	0.042***	0.043***	
	(0.019)	(0.013)	(0.012)	(0.015)	(0.013)	
$\log(Agr_{kt}/Agr_{jt})$	0.044***	0.028***	0.001	-0.041***	0.004	
-	(0.009)	(0.007)	(0.006)	(0.007)	(0.006)	
$\log(Cons_{kt}/Cons_{jt})$	0.125***	0.010	-0.027	-0.067**	-0.061**	
·	(0.038)	(0.027)	(0.025)	(0.030)	(0.025)	
$\log(Man_{kt}/Man_{jt})$	0.038*	0.010	0.020	-0.072***	0.072***	
	(0.022)	(0.016)	(0.015)	(0.017)	(0.015)	
$\log(\phi_{jk})$	-0.722***	-0.790***	-0.815***	-0.816***	-0.981***	
	(0.005)	(0.004)	(0.004)	(0.004)	(0.003)	
Nonparametric terms					·	
Spatio-temporal trends	Yes	Yes	Yes	Yes	Yes	
(ANOVA specification)						

With spatio-temporal trends, the displacments effects become even clearer, while the complementarity is not significant any longer...

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Endogeneity

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We might have endogenity issues due to omitted variables or reverse causality.

Example of reverse causality: insufficent flows of internal migrants in a given province induce local policy makers or local firms to expand the demand for international workers to compensate...

Lagging the explanatory variables one year partially mitigate reverse causality but might not be sufficient to eliminate it completely if migration variables are persistent over time.

To correct for endogeneity we need external instruments (i.e. variables correlated with the observed immigration flows but not with the error term). To this end, we follow Beine and Coulombe (2018) **control function approach** which consists in exploiting the information on the country of origin of foreign immigrants.



In brief (more details in the paper)... the control function is constructed in four steps:

- We estimate a negbin model that explains the magnitude of the flows of immigrants from each origin country *c* of the world to each province *k* for each year *t*;
- 2. Based on this estimation, we recover the predicted bilateral flows for each country-province-year triplet
- 3. We sum these up across country of origin to get a total predicted flow of immigrants by year and province and we use this as an instrument, i.e. the observed immigration rate is regressed against the predicted immigration rate in a regression model including all other exogenous variables included in the model before.
- 4. A smooth function of the residuals, **f(res)**, from this model is included in the original model to control for endogeneity.

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TABLE 4

Gravity models of internal mobility in Italy. Testing the relative incidence of foreign immigrants. Negative Binomial Estimates. Model specification (2) with spatio-temporal trends, accounting for endogeneity (second step CF approach)

		Italian citizens					
	Primary	Lower sec.	Upper sec.	Higher	residents		
Parametric terms		Coefficients (standard errors)					
Intercept	-2.486***	-0.861**	-1.328***	-0.954***	-2.557***		
	(0.484)	(0.398)	(0.322)	(0.304)	(0.221)		
$\log(Imm_{kt}/Imm_{jt})$	-0.402***	-0.141***	0.096**	0.548***	-0.298***		
	(0.073)	(0.053)	(0.049)	(0.059)	(0.051)		
$\log(Inc_{kt}/Inc_{jt})$	0.165**	0.044	0.079^{*}	0.057	0.099**		
	(0.068)	(0.049)	(0.046)	(0.055)	(0.047)		
$u_{kt} - u_{jt}$	-0.027***	-0.016***	-0.006**	0.011***	-0.030***		
J	(0.004)	(0.003)	(0.003)	(0.003)	(0.003)		
$\log(House_{kt}/House_{jt})$	0.007	0.002	-0.007	0.010	0.051***		
	(0.019)	(0.014)	(0.013)	(0.015)	(0.013)		
$\log(Agr_{kt}/Agr_{jt})$	0.045***	0.029***	0.001	-0.044***	0.004		
	(0.009)	(0.007)	(0.006)	(0.007)	(0.006)		
$\log(Cons_{kt}/Cons_{jt})$	0.151***	0.017	-0.031	-0.097***	-0.047*		
	(0.038)	(0.028)	(0.026)	(0.030)	(0.026)		
$\log(Man_{kt}/Man_{jt})$	0.017	0.003	0.025*	-0.036**	0.062***		
-	(0.022)	(0.017)	(0.015)	(0.017)	(0.015)		
$\log(\phi_{jk})$	-0.723***	-0.789***	-0.814***	-0.814***	-0.978***		
	(0.005)	(0.004)	(0.004)	(0.004)	(0.003)		
Nonparametric terms		χ^2 -test [EDF]					
f(res)	28.10***	12.28**	14.03***	104.39***	135.27***		
,	[2.163]	[3.096]	[3.040]	[3.215]	[3.956]		
Spatio-temporal trends	Yes	Yes	Yes	Yes	Yes		
(ANOVA specification)							

Controlling for endogenelity makes the results on displacement and complementarity effects by level of education of natives stronger.

The displacement effect on other foreign-born is also confirmed and stronger.

The instrument is strong F-stat 24165, p-value=0.000

Discussion and preliminary conclusions

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- Our results indicate, in the case of Italy, a displacement effect of the immigrants on the internal mobility of foreign-born residents and of Italian citizens with a low education level, but also a positive impact on the internal mobility of natives with a high education level
 - These findings suggest that interregional migration is an important mechanism through which the Italian labor market adjusts to immigration
 - They also open up an interesting question: why is the displacement only happening for low educated natives? Is it a **"pure" human capital story** (immigrants simply are all low educated) or is it also a **signaling problem** (immigrants cannot "signal" their skills even when they have them)?

Policy Implications

- On the one hand, incentivizing human capital accumulation is a viable option to reduce the displacement effect of immigration and, at the same time, to benefit low-skill workers (and the country overall) through increase in their productivity
- On the other hand, the most 'tricky' part is the integration of foreign immigrants and the identication of their skill level
- The missing ingredient in the integration process is a systematic assessment of the human capital of foreign immigrants. It is wrong to assume all of them have no or low skills.

- Where most (if not all) European countries are lagging behind is in mapping the skills, education and competences of foreign migrants
- The mapping and validation of these skills by the government should:
 - be included explicitly in the national dispersal policies to ensure a better matching between demand and supply of labor
 - help with the discrimination faced by humanitarian migrants with local employers often dismissing almost completely their qualications and/or work experience (Damos de Matos and Liebig, 2014)



alessandra.faggian@gssi.it



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• ANOVA-type decomposition (Lee and Durban, 2011): $f(x_{s1}, x_{s2}, x_{\tau}) = f_1(x_1) + f_2(x_2) + f_t(x_{\tau}) + f_{1,2}(x_1, x_2) + f_{1,\tau}(x_1, x_{\tau}) + f_{2,\tau}(x_2, x_{\tau}) + f_{12\tau}(x_1, x_2, x_{\tau})$

