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Original Scientific Paper

SUPPORTING IMMIGRATION FOR AN ENHANCED AGRICULTURAL SUSTAINABLE DEVELOPMENT WITHIN THE EUROPEAN UNION

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Abstract. Given the amplitude of migration in Europe and its potential to encourage or hinder the agricultural development, the general objective of this paper is to assess this possible potential on the long term, within a sustainable development frame of reference. Along these lines, we have built up several scenarios that focus on the agricultural results attained by ten EU Member States mainly targeted by immigrants, throughout the 2020-2025 period (sustainable development extrapolation). We have elaborated a set of indicators and within a panel in order to implement the spatial analysis and structural equation modelling (SEM), as methodological endeavour. The results obtained, verified by testing four hypotheses, show that a positive tendency in terms of increased government agri-innovation support is revealed on the long run, through the economic (labour) migration. Our findings outline that conclusive results of labour immigration could reverse generating unbalances in the agricultural sector. Thus, the need to develop accurate tailored policies is more than necessary by acknowledging the complex problems of the rural areas and those of international migration, as well as the major discrepancies among countries and stronger socio-economic interconnections.

Key words: agriculture sector, international migration, European integration, sustainable development, econometric modelling

JEL Classification: J61, Q10, C31

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1. INTRODUCTION

Around the world, agriculture represents an important sector, especially by considering its development potential (37% agricultural land out of the total land area in 2014, according to the World Bank, 2017). Still, "due to its dependence on natural factors" (Arisoy et al., 2017, p. 234), agriculture generates annual fluctuations of producers' income, due to the specific uncertainties (Cristea et al., 2008). This fact, in the long run, "leads to the migration of farmers" (Arisoy et al., 2017, p. 246), thus inducing complex implications upon the agriculture's results.

The global updated statistics reveal the magnitude of international migration, from which a high share of migrants are rural people, "revealed by the fact that around 40 percent of international remittances are sent to rural areas" (FAO, 2017a, p. 98).

A series of migration research (e.g. Card et al., 2010; Krause et al., 2016), and especially those focused on the Brexit implications after the June 2016 decision (e.g. Kierzenkowski et al., 2016), have mainly analysed its effects by considering various factors. However, the overall implications of migration, in Europe, on the agriculture field have been less analysed by specialists, although this sector has a considerable potential, mainly through the agricultural land. At the same time, there are a number of studies that have studied the impact of migration upon agriculture in North America (e.g. Barkley, 1990; Martin, 1993), Latin America (e.g. Balan, 1983; Aide & Grau, 2004) and Asia (e.g., Sunam, 2017). In Europe, however, there are few studies taking into account the migration implications upon the agricultural sector and most of these have outcomes for a particular country, such as Albania (Miluka et al., 2010). In a previous research into this field, we have investigated (Cristea & Noja, 2019) these complex implications at the level of the main targeted ten European Union (EU) countries, for the period 2000-2016.

Based on this framework, the general objective of this research is to forecast the implications of immigration upon the agriculture field on the long term (2020-2025 period), within the sustainable development goals (SDGs), for ten European Union Member States (MS) mainly targeted by immigrants (namely Germany, France, the UK, Austria, Sweden, Italy, Spain, Belgium, Denmark and Finland). Thus, this research represents a further enhancement of that made by Cristea and Noja (2019), on the fore of sustainable development. The SDGs are applied to all 193 countries of United Nations (UN), globally, without any distinction for "developed" or "developing" states. The SDGs are also centralized by the European Union (EU) Statistical Office (Eurostat database) for each country, both MS, and non-EU countries.

The next section of the paper comprises a synthesis of literature review, summarizing the main findings on migrants' implications upon the agricultural sector in different economies, along with the current situation of migration flows and main agricultural outcomes in the EU. Further are presented the data and the methodology applied, followed by main results, discussions and concluding remarks in the final stage of our research.

2. The Literature Review

2.1. Main findings of previous researches

There are several studies, which analyse the agriculture under the impact of the migration process around the world; still, in Europe, these potential implications did not represent a prevailing subject for researchers.

Thus, for *the North America* region, Barkley (1990) applied aggregate data for migration in the United States of America (USA) and "a two-sector model of occupational choice" (Barkley, 1990, p. 567) for the labour outputs in agriculture. He highlighted the importance of accurate policies for increasing the farms' income, with a further impact on the employment levels in agriculture. Martin (1993) studied the impacts of trade and migration upon agriculture in Mexico under the North American Free Trade Agreement (NAFTA). He pointed out that this kind of agreement between USA and Mexico would generate annual increase of migration from Mexico towards USA on the short-term by 100,000 persons. For the long-term, NAFTA is a keystone for reducing emigration from Mexico by enhancing the economic growth and better job creation in Mexico. The author found that the policies on the labour market of the hosting countries (e.g. USA), and also the regulation) would influence the migration process.

For Latin America, Balan (1983) investigated internal migration in relation with the social structure of agriculture for a period of 100 years, which, in the last stage, "led to the declining importance of peasant production and the increasing importance of wage labour" (Balan, 1983, p. 151). Studying migration in Latin American on the background of globalization, Aide and Grau (2004) have drawn up into attention the need for "social programs on preparing rural migrants for an urban environment and promoting ecosystem recovery in the marginal agriculture" (Aide & Grau, 2004, p. 1915). They underlined the impact of the urbanization process determined by rural-urban migration on the agricultural field, especially, on ecosystem conservation.

For Asia, Sunam (2017) highlighted the fact that for some researchers and policymakers the path to prosperity would be the non-agrarian sectors, while others consider land of major importance, especially, in poverty decreasing. He criticizes the narrow framework of their analysis, which overtakes the "strong linkages between agriculture, migration and rural labour, but also stay silent on how rural people interpret changes or continuities in their livelihoods" (Sunam, 2017, p. 67).

Working in the agricultural area, "is associated with low and insecure incomes, poor occupational safety and health conditions, gender inequality in pay and opportunities, and limited access to social protection" (FAO, 2017a, p. 100). Also, Barone and Mocetti (2011) attest that low-skilled migration tends to grow participation among native women to labour market, thus generating the increase of productivity and output in the agricultural sector. Therefore, Boswell (2016) underlines the necessity for better "*training and education policies (ALMPs)*" for natives (e.g. British labour force) since a significant reduction of labour will generate damaging effects, including in the agricultural sector. Card et al. (2010) identified and evaluated another variable, "*the passive labour market policies (PLMPs)*", with major effects over the labour market, which refer mainly to: keeping the income for "the unemployment period through unemployment benefits; the earlier retirement which

facilitate total or partial retirement of elder workers with decreased possibilities in finding a job" (Cristea & Noja, 2019, p. 120).

Regarding migration, since most policy-makers from developing countries tried to diminish it, or even to ignore it, they recommend to sustain migration, "becoming aware of its potential", by following priorities: "reducing the costs and risks faced by migrants; ensuring that entitlements to state services are portable; facilitating remittances; improving accountability and transparency in labour markets; and raising awareness of labourers' rights" (Wiggins & Deshingkar, 2007, p. 3). In Europe, the Brexit decision (being in process of finalization at the end of October 2019) will have a significant impact upon migration flows. A possible diminishing of migration inflows could generate negative effects (especially for the UK), particularly for low-salary fields, mainly targeted by migrants, such as manufacturing, care jobs, food processing (Rienzo & Vargas-Silva, 2012).

2.2. Current situation of agriculture and international migration within the EU

In 2015, within EU-28, *the total utilized agricultural area* was of 43% of EU land area, and comprised the following components: *arable land*, 59.74% of the total utilized area; *permanent grassland*, 33.36%; *permanent crops*, 6.55%; "*other agricultural land such as kitchen gardens* (small areas of total utilized agricultural area)" (European Commission, 2018). The highest share of *the total utilized agricultural area* (Figure 1a) is owned by France (16.06% of the EU-28), Spain (13.18%), the UK (9.46%), Germany (9.23%), Poland (7.94%), Romania (7.64%) and Italy (6.98%) (European Commission, 2018). The same countries accounted for the most significant *arable land* within the EU, in the following order: France, Spain, Germany, Poland, Romania, Italy and the UK (Figure 1b).



Fig 1 Utilized agricultural area (a) and arable land (b) in EU, 2015 *Source:* Authors' processed in Stata, Eurostat data

However, although the share of agricultural land in the EU is quite high (above the average worldwide level of 37%) with considerable potential alongside farms and fisheries,

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the agriculture's share in EU's GDP is very low, representing an average of 2.32% in 2016, compared to 4.26% in all EU MS, respectively 4%, globally (World Bank, 2017). Thus, we can notice that, on the one hand, among the countries with high agricultural potential there are mostly developed countries (France, Spain, the UK, Germany, Italy), even though the share of agriculture in these countries' GDP is not significant (1% of GDP for the UK and Germany, 2% for France and Italy, and 3% for Spain) (World Bank, 2017). On the other hand, Poland and Romania have a large share of agricultural area among EU-28 (ranks 5 and 6), being the most important agricultural producers from the EU-28 (Marcu et al., 2015; Done et al., 2012). Furthermore, these two countries face large emigration flows, with important negative spillovers "on the size and structure of the labour force" (Noja et al., 2018, p. 3), particularly in agriculture. Moreover, the highest contribution of agriculture to GDP among the EU countries is for: Bulgaria (5%), Croatia, Greece, Hungary, Latvia, Romania and Slovak Republic (4% of GDP in each country). These countries are mainly emigration or labour-exporting countries so as the migration process reflects negatively in terms of long-term economic development.

Nowadays, agriculture represents "the world's biggest employer and largest economic sector for many countries. Yet rural people – who produce 80 percent of our food – make up four-fifths of the global poor" (FAO, 2017b, p. 5). Thus, if at the global level, *the employees in agriculture* registered a share of 26.7% (866.3 millions) in the total number of employees in 2016 (3240 millions), for EU-28 this share was 4.3% in 2016 (9,771 thousand employees out of 226,716 thousand overall) found in a decreasing trend from 7.9% in 2000 (ILO, 2017). By comparing Eastern with Western Europe, the largest share of agricultural workers can be found in Eastern Europe with 9.4% of total employees, compared to 2.1% in Western Europe (Figure 2).



Fig 2 Shares of employment in agriculture in EU-28, 2016 Source: Own process data offered by the International Labor Organization (ILO)

Regarding *the international migration situation*, in 2015, large shares of international migrants were accounted for by Europe (over 31%), followed by Asia (up to 31%) (UN DESA, 2015). The total number of migrants in Europe (over 76 million people) has increased with almost 4 million people compared to 2014, representing 31.25% from the total stock of migrants at the worldwide level (over 243 million persons) (UN DESA, 2015). The highest share of migrants' stock in Europe is in Western Europe (35.96%), followed by Eastern

Europe (25.85%), encompassing mostly rural people, "revealed by the fact that around 40 percent of international remittances are sent to rural areas" (FAO, 2017a, p. 98).

The most targeted 10 EU countries by migrants in 2015 were: "Germany (12 million migrants stock), the UK (8.5 million), France (7.8 million), Spain (5.9 million), Italy (5.8 million), Sweden (1.6 million), Austria (1.5 million), Belgium (1.4 million), Denmark (572 thousands), Finland (315 thousands)" (Noja et al., 2018, p. 4). Bilaterally, the main migrant sending economies are in fact developing countries, with an important rural field into the economy and a high share of agricultural employees looking for cross-border low-skilled jobs (e.g. Poland, with 3.6 million of migrants, Romania, almost 3 million).

The inflow trend in 2016 compared to 2015 increases into the following EU-10 countries: Germany (by 30%), France (by 4%), Spain (by 18%), Sweden (by 26%), Austria (by 3%), Belgium (by 7%) and Finland (by 8%). The decreasing is registered into the UK (by -6%), Italy (by -21%), and Denmark (by -5%) (OECD, 2017). At the end of 2016, the refugees' situation registered "a total of 22.5 million displaced people, being considered the highest on record, although the annual rate of growth has slowed since 2012" (OECD, 2017, p. 31). Large shares of asylum seekers in Europe were registered by Germany and Italy, "Germany remained the top recipient, with over 720,000 applications, followed by the United States (262,000) and Italy (123,000)" (OECD, 2017, p. 32).

Thus, we can see a double trend in international migration, one regarding the *labour mobility* (migrants searching for better jobs), and the other, the flow of refugees and asylum applicants (called *"humanitarian migration"*). On these credentials reflecting the migration amplitude and agriculture potential in Europe, along with the Brexit phenomenon and attaining the SDGs by all countries, we further investigate our general and specific research objectives.

3. DATA AND METHODOLOGY

3.1. Data used

In our investigation, we analyse the data on the long term (2020-2025), within a sustainable development frame of reference, especially after the Brexit negotiation deadline (October 2019). Thus, we take into account the following dimensions for our data (variables), most of them included into the SDGs: *reducing poverty* (SDG1, "No Poverty"), highly connected with the agricultural sector; *well-being* (SDG3) through the life expectancy at birth; *quality of education* (SDG4), by educational attainment for tertiary education; *decent work* (SDG8) for which we measure labour market indicators, such as employment and unemployment rate, the labour market policies (ALMPs and PLMPs respectively); *industry innovation implications* (SDG9), by the research and development (RD) expenses for the business enterprise sector; *income and living conditions*, measured through net earnings. These dimensions are included into 3 categories of data, representing: *a) the main agricultural outcomes*; *b) the immigration situation*; *c) SDGs* presented above and other *representative indicators for the international activity*, namely: Foreign Direct Investment (FDI) (UNCTAD, 2017), and KOF index of globalization (KOF) (ETH Swiss Federal Institute of Technology Zurich, 2017).

Thus, for the agricultural outcomes, we've considered the following data: "Value Added by kind of economic activity (VA_AGRI), including agriculture, hunting, forestry,

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fishing" (UNCTAD, 2017); "agricultural factor income per annual work unit (AGRI_AWU); government support to agricultural RD (GOV_AGRI)" (European Commission, 2018); "merchandise trade matrix - food group (X_FOOD); merchandise trade matrix - agricultural raw materials (X_AGRI_RM)" (UNCTAD, 2017). *The immigration situation* is measured by "flows of immigrants and foreign population (IMIG)" (European Commission, 2018; OECD, 2017), and "flows of asylum applicants" (ASYL) (UNHCR, 2017). *For SDGs attainment* (European Commission, 2018), we include in our research the following indicators: "at-risk-of-poverty rate" (POV); "life expectancy at birth" (LE); "educational level reflected through the educational attainment for tertiary education" (EDU_T); "employment rate" (ER); "unemployment rate of foreign population" (UR_F); "the active labour market policies" (ALMPs) and "the passive labour market policies" (PLMPs); "RD expenditures for the business enterprise sector" (BERD); "annual net earnings of a twoearner married couple with two children" (EARN).

The panel includes ten EU countries having highest immigration flows, analysed during 2020-2025 in order to reveal the long run effects (sustainable development), as extrapolation series based on statistical data for 2000-2016.



Fig 3 AGRI_AWU trends in EU-10 migrant receiving countries, during 2005-2015 Source: authors' process of Eurostat data

VA_AGRI, in 2015, is at the highest level in Italy and Spain (UNCTAD, 2017). AGRI_AWU appraises the productivity for the agriculture field, by counting "the income generated by farming, which is used to remunerate borrowed or rented factors of production (capital, wages and land rents) as well as own production factors (own labour, capital and land)" (European Commission, 2018). Annual work unit (AWU) represents "full-time equivalent employment (corresponding to the number of full-time equivalent jobs), i.e. as total hours worked divided by the average annual number of hours worked in full-time jobs within the economic territory" (European Commission, 2018). AGRI_AWU has registered extremely high levels in Denmark, especially after the global economic crisis in 2009, reaching a peak of 60,716 euro in 2012, well above the ones accounted by the other countries considered. However, Denmark currently faces major problems in this respect since AGRI_AWU has fallen severely starting with 2014 (Figure 3). Relatively constant high levels are registered in France, Belgium and Sweden during the entire time period.

GOV_AGRI data have registered a major increase in Germany after 2004, this country following an upward path afterwards, well above the one accounted by the other EU countries considered. At the same time, Spain has made important efforts to support RD in agriculture, with important positive output during 2004-2008 period. However, after the global crisis in 2008-2009 this support has significantly diminished, even though it is still above the one performed by most EU economies (Fig 4). France registered important contribution in RD for agriculture in 2016, and the UK saw a visible decrease in the amounts allocated for RD in agriculture.



Fig 4 GOV_AGRI in EU-10 migrant receiving countries during 2004-2016 Source: authors' process of Eurostat data

X_FOOD and X_AGRI_RM capture "the international trade performance of the agricultural field" (Cristea & Noja, 2019, p. 117) in selected countries and reveal the importance of this sector for EU-10 economies. These two indicators highlight the dominant position for Spain (food) and Austria (raw materials) (UNCTAD, 2017).

3.2. Methodology applied into research

Based on our general objective and by reviewing the current state of the literature on agricultural results under the immigration impact, we have set out the following *research hypotheses* (*H*):

- H1. There is a strong relationship between VA_AGRI and IMIG/ASYL;
- H2. There is a strong relationship between AGRI_AWU and IMIG/ASYL;
- H3. IMIG and ASYL induce increased GOV_AGRI;
- *H4. IMIG and ASYL improve* X_*FOOD and* X_*AGRI_RM.*

First, we have applied *the standardisation procedure* on our data, for a proper equivalence among EU-10 MS, and to remove the fluctuation and associated disparities within the panel, according to the Eq. (1) (OECD, 2005):

$$y_i = \frac{x_i - mean}{sd} , \qquad (1)$$

where: y_i represents the composite indicators, x_i is the crude value of the indicator; and *sd*, the standard deviation".

To forecast the variables during 2020-2025 we have applied *the linear extrapolation*, based on statistical time series for 2000-2016. "The extrapolation formula used is described by Eq. (2).

$$y(x) = y_1 + \frac{x - x_1}{x_2 - x_1} (y_2 - y_1),$$
(2)

where: x_1 , y_1 and x_2 , y_2 are the two endpoints of a linear graph; and x represents the point of which value is to be extrapolated" (Noja et al., 2018, p. 6).

Then, we have applied *particular macro-econometric models* in order to appraise the effects lead by immigration flows upon receiving economies' agricultural results, namely, *spatial analysis models* (lag and error), Eq. (3a) and (3b) (Viton, 2010).

• *"Spatial lag* models:

$$y = \lambda W y + X \beta + u, \tag{3a}$$

• *Spatial error* models:

$$y = X\beta + u, \ u = pWu + v, \tag{3b}$$

where: W is the inverse distance weights matrix".

The standard macro-econometric model for VA_AGRI as dependent variable is built up as "*a baseline panel regression model*, *but reconfigured through the spatial procedures*" (Eq. 4a) for lag model, and Eq. 4b for error model):

$$VA_AGRI_{ii} = \lambda WVA_AGRI_{ii} + \beta_0 + \beta_1 IMIG_{ii} + \beta_2 ASYL_{ii} + \beta_3 ER_{ii} + \beta_4 BERD_{ii} + \beta_5 ALMPs_{ii} + \beta_6 EDU_T_{ii} + \beta_7 EARN_{ii} + \beta_8 FDI_I_{ii} + \beta_9 FDI_O_{ii} + u_{ii},$$
(4a)

$$VA_AGRI_{it} = \beta_0 + \beta_1 IMIG_{it} + \beta_2 ASYL_{it} + \beta_3 ER_{it} + \beta_4 BERD_{it} + \beta_5 ALMPs_{it} + \beta_6 EDU_T_{it} + \beta_7 EARN_{it} + \beta_8 FDI_I_{it} + \beta_9 FDI_O_{it} + \rho Wu_{it} + \upsilon_{it}.$$
(4b)

The final stage of our methodology consists in applying the *structural equations modelling (SEM)*. SEM reveals the integrated analysis of immigration interlinkages (direct, indirect and total) with agriculture field. The general representation of SEM is shown into "equation system (5).

$$\begin{cases} b_{11}y_{2t} + \dots + b_{1m}y_{mt} + c_{11}x_{1t} + \dots + c_{1n}x_{nt} = \varepsilon_{1t} \\ b_{21}y_{2t} + \dots + b_{2m}y_{mt} + c_{21}x_{1t} + \dots + c_{2n}x_{nt} = \varepsilon_{2t} \\ \dots \\ b_{m1}y_{1t} + \dots + b_{mm}y_{mt} + c_{m1}x_{nt} + \dots + c_{mn}x_{nt} = \varepsilon_{mt} \end{cases}$$
(5)

where: *t* is the number of time periods; b_{ij} represents the y_{ij} endogenous variable's parameters; c_{ij} are the x_{ij} exogenous variable's parameters, i=1, ..., m; j=1, ..., n'' (Noja et al., 2018, p. 8).

4. RESULTS AND DISCUSSIONS

4.1. Variables extrapolation

Figure 5 reveals the 2025 timeline forecast of the agricultural indicators included in our analysis (the sustainable development frame) for EU-10 countries. Thus, VA_AGRI is at the highest level in the UK and Belgium for 2025 (Figure 5a). These forecasts shed initial

lights on the importance of agriculture for the UK's economy after the major challenges encountered within the Brexit context. Relatively constant high levels for AGRI_AWU are registered in France and Belgium (the same tendency as in 2015) (Figure 5b). Considering GOV_AGRI, the UK saw a visible decrease in the amounts allocated for RD in agriculture due to Brexit (Figure 5c). High levels of RD in agriculture will be registered by France and Germany. X_FOOD will tend to increase in Germany and Spain (Figure 5d). X_AGRI_RM of the UK will tend to decrease considerably until 2025 (Figure 5e).



Source: authors' process in Stata

4.2. Spatial analysis models

Spatial results (Table 1) show that the immigration process requires a particular attention from policy makers across Europe, since our estimations for the 2020-2025 period reveal that the positive outcomes of IMIG could reverse leading to unbalances even in the agricultural sector (the statistically significant estimated coefficient is – 0.0569). This is also the case of humanitarian migration (ASYL), since the results point out that increased inflows of ASYL tend to rather induce negative spillovers upon the VA_AGRI (the estimated coefficients are -0.111 and -0.131).

The uncertainty brought after 2020 reflects negatively also considering *FDI inwards*, *which generate a slight diminishing in the agricultural results*. Negative impacts are also registered for the ALMPs, and EDU_T, similar with those obtained for the period 2000-

2016 (Cristea & Noja, 2019). In contrast, beneficial results can be registered in case of *FDI outwards, with positive impacts on the agriculture's productivity,* thus raising the contribution of agriculture to GDP. Positive effects will be registered in terms of ER and EARN. Moreover, the results highlight *the significance of BERD in growing the VA_AGRI* (the estimated coefficients are positive and extremely significant at a threshold of 1%).

	(1)	(2)	(3)	(4)	
Variables	Spatial_lag	Spatial_lag	Spatial_error	Spatial_error	
	(robust)	(bootstrap)	(robust)	(bootstrap)	
IMIG_st	-0.0354	-0.0354	-0.0569*	-0.0569*	
	(0.0244)	(0.0310)	(0.0242)	(0.0275)	
ASYL_st	-0.111****	-0.111***	-0.131***	-0.131***	
	(0.0164)	(0.0168)	(0.0155)	(0.0174)	
ER_st	0.273^{**}	0.273^{*}	0.286^{**}	0.286^{*}	
	(0.0839)	(0.135)	(0.0875)	(0.114)	
BERD_st	0.337***	0.337^{*}	0.362^{***}	0.362^{**}	
	(0.0904)	(0.148)	(0.0941)	(0.126)	
ALMPs_st	-0.182***	-0.182***	-0.219***	-0.219***	
	(0.0249)	(0.0178)	(0.0167)	(0.0230)	
EDU_T_st	-0.0870***	-0.0870***	-0.0919***	-0.0919***	
	(0.0182)	(0.0296)	(0.0192)	(0.0248)	
EARN_st	0.0923***	0.0923*	0.102^{***}	0.102**	
	(0.0258)	(0.0407)	(0.0285)	(0.0326)	
FDI_I_st	-0.0600***	-0.0600*	-0.0550***	-0.0550*	
	(0.0197)	(0.0250)	(0.0212)	(0.0277)	
FDI O st	0.0139	0.0139	0.0163	0.0163	
	(0.0329)	(0.0368)	(0.0344)	(0.0435)	
cons	0.431	0.431	0.0426	0.0426	
_	(0.236)	(0.299)	(0.143)	(0.157)	
Rho	. ,	. ,	· · · ·	. ,	
cons	0.603^{*}	0.603**			
	(0.304)	(0.227)			
Sigma					
cons	0.220^{***}	0.220^{***}	0.223^{***}	0.223^{***}	
	(0.0241)	(0.0203)	(0.0253)	(0.0242)	
Lambda	(0.02.12)	(0.010200)	(0.0200)	(***= !=)	
cons			0.562	0 562	
_00115			(0.381)	(0.719)	
Test parameters	chi2(10)-141.29	chi2(10) - 602.05	$\frac{(0.301)}{(10)-675.34}$	$\frac{(0.717)}{\text{chi}^2(10) - 753.39}$	
rest parameters	n=0.0000	n=0.0000	n=0.0000	n=0.0000	
IM	1 800	1 800	0.040	0.040	
	(0.168)	(0.168)	(0.330)	(0.330)	
Wold tost	2 026	2 026	2 174	2 174	
of rho/lambda	(0.047)	(0.047)	2.174	2.174	
A agontable range for the	(0.047)	(0.047)	(0.140)	(0.140)	
Acceptable range for limit day $2.500 < 1.000$;					
Acceptable range for familia $-5.509 \le 1000$ Moreov's LIMIC 1-0.420, $= -0.000, ASVI 1-0.241, = -0.000, VA ACDI-0.254, = 0.000$					
woran's i iwiiG I=0.4	29; p=0.000; ASY	L 1=0.241; p=0.00	$\frac{10}{10}$; $vA_AGKI=0.23$	94; p=0.000	
N	60	60	60	60	

Table 1 Spatial lag and error models (robust and bootstrap) for VA_AGRI, 2020-2025

Note: standard errors in parentheses; ${}^{*}p < 0.05$, ${}^{**}p < 0.01$, ${}^{***}p < 0.001$ *Source:* authors' research in Stata These aspects were also underlined by Kierzenkowski et al. (2016), Jirasek (2017) and Cristea and Noja (2019), which proved that a reorient to innovation strategies combined with supporting the international trade activity, could overcome the potential difficulties registered after the Brexit.

The spatial lag models show that the positive evolution based on increased GOV_AGRI, X_FOOD and X_AGRI_RM is registered on the long-term (Table 2), but only under the influence of IMIG. For ASYL, the negative coefficients highlight the significance of "accurate strategies, policies and targeted measures to be applied by the major receiving countries to manage with growth immigration inflows and to properly cope with the refugee crisis in Europe" (Cristea & Noja, 2019, p. 118). The immigrants can be well integrated into receiving countries and can therefore increase the agricultural and overall outcomes, as reflected by our estimations, given an accurate harmonized strategy applied at the European level.

Variables	(1)	(2)	(3)	(4)
	AGRI_AWU_st	GOV_AGRI_st	X_FOOD_st	X_AGRI_RM_st
IMIG_st	0.0458	0.311****	0.486^{***}	0.280^{***}
	(0.0349)	(0.0275)	(0.0294)	(0.0461)
ASYL_st	-0.379***	-0.0835***	-0.0288^{*}	-0.0335***
	(0.0408)	(0.0110)	(0.0127)	(0.0120)
ER_st	0.0606	-0.531***	-0.410***	-0.0639
	(0.148)	(0.0855)	(0.0806)	(0.0745)
BERD_e_st	0.483^{**}	0.709^{***}	-0.0630	0.213**
	(0.148)	(0.0757)	(0.0746)	(0.0647)
ALMPs_st	-1.144***	0.115^{***}	0.321^{***}	-0.190***
	(0.102)	(0.0283)	(0.0191)	(0.0371)
EDU_T_st	-0.0359	-0.131***	-0.127***	0.0305^{*}
	(0.0261)	(0.0160)	(0.0159)	(0.0135)
EARN_st	-0.171***	-0.0612**	-0.374***	-0.323***
	(0.0640)	(0.0226)	(0.0285)	(0.0513)
FDI_I_st	0.296^{***}	0.102^{***}	-0.110***	0.0420^{**}
	(0.0404)	(0.0189)	(0.0160)	(0.0134)
FDI_O_st	-0.203****	-0.371***	0.352^{***}	-0.0504**
	(0.0507)	(0.0282)	(0.0219)	(0.0166)
_cons	2.089^{***}	1.031****	1.162***	0.278^{**}
	(0.201)	(0.0986)	(0.114)	(0.107)
Rho				
_cons	-0.0122	0.899^{***}	-0.135	0.116
	(0.260)	(0.0992)	(0.132)	(0.426)
Sigma	de de de		de de de	ded d
_cons	0.465^{***}	0.253^{***}	0.202^{***}	0.197^{***}
	(0.0383)	(0.0244)	(0.0226)	(0.0197)
Ν	60	60	60	60

Table 2 Results of Spatial lag models, 2020-2025

Note: Standard errors in parentheses; ${}^{*}p < 0.05$, ${}^{**}p < 0.01$, ${}^{***}p < 0.001$ *Source*: authors' research in Stata

Based on spatial lag models, we can say that: *H1. There is a strong relationship between* VA_AGRI and IMIG/ ASYL is partially fulfilled (only for ASYL); H2. There is a strong relationship between AGRI_AWU and IMIG/ ASYL is partially fulfilled (only for ASYL);

H3. IMIG and ASYL induce increased GOV_AGRI is partially fulfilled (only for IMIG); H4. IMIG and ASYL improve X_FOOD and X_AGRI_RM is partially fulfilled (only for IMIG).

4.3. SEM models

Considering SEM, we have verified Wald tests for equations, compared Likelihood Ratios (LR) outcomes (LR test for model versus saturated and baseline versus saturated) and Information criteria (Akaike's, Bayesian). Thus, we have obtained 5 models (VA_AGRI, AGRI_AWU, GOV_AGRI, X_FOOD and X_AGRI_RM) with conclusive results for 2020-2025 (Figure 6).



Source: authors' process in Stata

Thus, in the 2020-2025 sample, we acknowledge that the extrapolation procedure is subject to increasing uncertainty and the results obtained must be interpreted with caution. Still, also in this case, the major determinants of the migration decision are the employment opportunities for the foreign population, and the living standards at destination reflected here particularly through the earnings levels (0.32 estimated coefficient associated with the EARN variable). The results are opposite to Rienzo and Vargas-Silva (2012), which revealed a reversed connection between migration flows and migrant's earnings (especially for the UK), but also to Cristea and Noja (2019), revealing negative implications for the period 2000-2016. Moreover, the total employment rate (ER) increases significantly under the labour immigration effects (extremely statistically significant estimated coefficients for the IMIG variable, 0.22; while the ASYL coefficients are negative, but statistically insignificant). The further impact of all these implications upon the agricultural sector for the 10-EU MS is positive only in terms of increased GOV_AGRI, opposite to those obtained for the period 2000-2016 (Cristea & Noja, 2019). However, the estimations reveal that there is evidence to attest that an unfavourable implication on agricultural productivity (AGRI_AWU has a negative estimated coefficient of -0.71^{***} that is extremely statistically significant) at on this sectors' contribution to GDP (VA AGRI estimated coefficient is - 0.18^* significant at the 0.05 level), if the immigration process is not properly managed in the Brexit framework (deadline negotiations 2019). Moreover, there are also negative effects upon the international trade activity deployed in the agricultural sector as reflected through a reduction both in X_FOOD (-0.22 less statistically significant) and X_AGRI_RM (-0.24^{*} significant at the 0.05 level) in the light of new trade agreements established after the Brexit.

Thus, considering the SEM models, H1, H2, H3 are fulfilled, and H4 is rejected.

5. CONCLUSIONS

The results obtained, verified by testing 4 hypotheses (through both econometric modelling procedures applied, spatial analysis and SEM, respectively), highlight: a strong direct linkage between the value added by the agricultural sector and the immigration flows (mainly humanitarian migration) (H1); instead, the productivity in the agricultural sector (measured through the agricultural factor income per AWU) is affected (negatively) only by humanitarian migration, labour migration having no statistical significance (low intensity) (H2); immigration flows (economic migration) induce increased governmental efforts focused on research and development to sustain the agricultural sector (H3); immigration flows (economic migration) improve the international trade outcomes for the agricultural sector (basic food and raw materials) (H4).

Thus, the need to develop policies for building a *new Europe* is more than necessary by acknowledging the complex problems of the rural areas and those of international migration, especially on the background of Brexit decision, characterized by the redemption of major discrepancies and stronger socio-economic connections.

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PODRŠKA IMIGRACIJI ZA POBOLJŠANJE POLJOPRIVREDNOG ODRŽIVOG RAZVOJA UNUTAR EVROPSKE UNIJE

Uzevši u obzir veličinu migracija u Evropi i njihov potencijal da poboljšaju ili unazade poljoprivredni razvoj, opšti cilj ovog rada je da proceni njihove potencijalne dugoročne posledice, u okviru referentnog okvira održivog razvoja. S tim u vezi, razradili smo nekoliko scenarija koji se fokusiraju na poljoprivredni razvoj deset zemalja-članica EU sa najvećim prilivom imigranata, u period 2020-2025 (ekstrapolacija održivog razvoja). Razradili smo set pokazatelja da bismo implementirali prostornu analizu i modeliranje strukturalnih jednačina (SEM), kao metodološki napor. Dobijeni rezultati, verifikovani tertiranjem četiri hipoteze, pokazuju da se na duge staze otkriva pozitivna tendencija u smislu povećane državne podrške agri-inovacijama, kroz ekomomsku migraciju (migraciju radne snage). Naši nalazi u glavnim crtama daju da bi konačni rezultati migracije radne snage mogli da promene stvaranje disbalansa u poljoprivrednom sektoru. Stoga, potreba da se razviju precizne prilagođene politike je više nego potrebna, time što će se prepoznati kompleksni problem ruralnih oblast ii međunarodnih migracija, kao i velike razlike među zemljama i veća socio-ekonomska povezanost.

Ključne reči: sector poljoprivrede, međunarodne migracije, Evopske integracije, održivi razvoj, ekonomski modelling