The Broadband Penetration in Europe

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Abstract

In this article we estimate the determinants of broadband penetration in Europe. We use data from the European Innovation Scoreboard of the European Commission for 37 countries in the period 2010-2019. We apply Panel Data with Fixed Effects, Panel Data with Random Effects, WLS, OLS and Dynamic Panel. We found that the level of "Broadband Penetration" in Europe is positively associated to "Enterprises Providing ICT Training", "Innovative Sales Share", "Intellectual Assets", "Knowledge-Intensive Service Exports", "Turnover Share SMEs", "Innovation Friendly Environment" and negatively associated with "Government procurement of advanced technology products", "Sales Impact", "Firm Investments", "Opportunity-Driven Entrepreneurship", "Most Cited Publications", "Rule of Law". In adjunct we perform a clusterization with k-Means algorithm optimized with the Silhouette Coefficient and we find the presence of three different clusters. Finally, we apply eight machine learning algorithms to predict the level of "Broadband Penetration" in Europe and we find that the Polynomial Regression algorithm is the best predictor and that the level of the variable is expected to increase of 10,4%.

Keywords: innovation and invention; processes and incentives; management of technological innovation; R&D; technological change; intellectual property; intellectual capital.

JEL Classification: O30; O31; O32; O33; O34.

Introduction

Introduction

In this article we investigate the determinants of broadband penetration in 37 European countries in the period 2010 - 2019. Broadband penetration is essential to promote the effective digitalization of the economy. In effect broadband penetration can sustain Internet of Things, telemedicine, smart cities and the promises of artificial intelligence and machine learning.

The ability to promote innovation technology especially in ICT sector is deeply associated with the efficiency and capacity of broadband penetration. Innovation technology is a driving force of economic growth such as in the Solow's model (Solow 1956, 65-94) in the theory of endogenous growth (Romer 1994, 3-22) and in the case of Schumpeterian economics (Schumpeter 1934). Furthermore, innovation technology has positive effects on human resources (Leogrande and Costantiello 2021, 240-259) and on the ability of firms to improve their sales either at an international level (Costantiello, Laureti, De Cristoforo and Leogrande 2021, 19). The investment in innovation technology is also positively associated to deeper expenditures in Research and Development either at a country level (Leogrande, Massaro and Galiano 2020, 72-101) either in the corporate dimension (Leogrande, Massaro and Galiano 2020, 186-201). Finally, innovation technology, in the sense of digitalization, also has positive effects on employment (Costantiello and Leogrande 2020, 166-187). For all these motivations it is important to analyze the question of broadband penetration for its role in promoting digitalization and innovation technology. Broadband penetration requires investments in fiber optics (Agboje, Adedoyin and Ndujiuba 2017, 1-12). There is a positive relationship between broadband penetration and the economic growth in Asia either in high income countries either

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in medium income countries (Kurniawati 2021), (Alam, Sultana and Rayhan 2019). The development of broadband penetration infrastructures improves total factor productivity and GDP growth (Kabaklarli and Atasoy 2019, 105-114). The investments in internet penetration is positively associated to economic development and growth in Africa (Kabaklarli and Atasoy 2019, 105-114). The creation of political economies that are specifically oriented to the creation of digital infrastructures in developing countries, such as in the case of Indonesia, shows that the effect on trade and ecommerce is structurally higher than the effect of human capital in terms of HDI (David and Grobler 2020, 1394-1418).

The development of telecommunication infrastructure also has a positive role in promoting digitalization and economic growth especially in high performing EU-OECD economies (Oztunc, Demirbas and Orhan 2019, 1-16). The investment in broadband penetration in Asian countries is positively associated to GDP growth especially in the case of low inflation as a tool to promote job creation in the digital economy (Shodiev, Turayey and Shodiyev 2021). While on one side, families and individual customers could prefer mobile internet connection to broadband penetration, since these groups use internet for communication and entertainment, on the other side firms and business activities express a demand of broadband penetration to improve the profitability of their activities (Dwiardi 2020, 41-58). Broadband penetration has negative impact on social capital in terms of relationships and interactions since users tend to prefer virtual life to real life (Geraci, Nardotto, Reggiani and Sabatini 2018).

The diffusion of broadband penetration is positively associated to population density (Benseny, J., Töyli, J., Hämmäinen, and Arcia-Moret 2019, 139-155) Broadband penetration can have a positive effect on productivity and regional convergence such as in the case of Brazil (Jung and López-Bazo 2020). The investment in broadband penetration has positive effect on economic growth at regional level in Russia even in the theoretical and applicative framework of the Solow's model (Imasheva and Kramin 2020, 26-28). The mechanism of pricing can have a role in shaping the degree of broadband penetration among countries *i.e.* tariff diversity is positively associated to a widespread broadband penetration (Lange 2017, 285-312). The diffusion of broadband penetration improves the ability of citizens to actively participate in political debate in Europe (Hendriks Vettehen, Troost, Boerboom, Steijaert and Scheepers 2020, 967-987).

Since the diffusion of internet at a country level is positively associated to the number of customers and the quantity of digital contents then countries that are interested in promoting broadband penetration should also improve internet services (Na, Hwang and Kim 2020, 97-111). The prediction of broadband penetration in developing countries such as India can also offer an outlook on the ability to promote economic growth and an improvement in Human Development Index-HDI (Chattopadhyay and Majumdar 2019, 110-119). There also controversial relationship among broadband penetration, income inequality and digital divide in the sense that either income inequality can reduce broadband penetration and either that digital divide can weaken the relation between digitalization and economic growth (Aissaoui 2017, 799-808).

The impact of telecommunication infrastructure on economic growth is greater for developing countries in respect to developed ones (Maneejuk and Yamaka 2020). Positive relationships between broadband penetration and economic growth are present also in Latin America (Alderete 2017, 549-569). Similar relationships also characterize the Nigerian economy (Akinwale, Sanusi and Surujlal 2018, 129-142) and Middle East countries (Habibi and Zabardast 2020). Broadband penetration in South Korea has reached highest levels as a consequence of a co-opetition between public sector and the market and the presence of firms able to offer high qualitative services to customers that show a significative orientation towards product and service innovativeness (Shin and Koh 2017, 31-38). Empirical studies have shown that an increase in 10% of mobile broadband penetration can rise GDP per capita of an amount in the range 0,59-0,76% at a country level if the economy also have a developed industrial sector and the human capital necessary to create value added through digitalization (Bahia, Castells and Pedrós 2019). The article continues as follows: the second paragraph presents the econometric model, the third paragraph contains the cluster analysis, the fourth paragraph shows the results of the machine learning algorithms used to predict the value of broadband penetration, the fifth paragraph concludes.

1. The Econometric Model

We use data from European Innovation Scoreboard in the period 2010-2019 for 37 countries to estimate the following equation:

BroadbandPenetration,,

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= a_1 \\ + b_1(Government \ procurement \ of \ advanced \ technology \ products)_{it} \\ + b_2(Firm \ investments)_{it} + b_3(Enterprises \ providing \ ICT \ training)_{it} \\ + b_4(Innovation - friendly \ environment)_{it} \\ + b_5(Innovative \ sales \ share)_{it} + b_6(Intellectual \ assets)_{it} \\ + b_7(Knowledge - intensive \ services \ exports)_{it} \\ + b_8(Most - cited \ publications)_{it} \\ + b_9(Opportunity - driven \ entrepreneurship)_{it} + b_{10}(Rule \ of \ law)_{it} \\ + b_{11}(Sales \ impacts)_{it} + b_{12}(Turnover \ share \ SMEs)_{it} 
 (1)
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where i = 37 and t = [2010, 2019].

The level of broadband penetration is defined as a relationship between a numerator that is the number of enterprises with a maximum contracted download speed of the fastest fixed internet connection of at least 100 Mb and a denominator that is based on the total number of enterprises. The level of broadband penetration is a measure that capture the ability of European countries to develop a digital economy based on e-commerce and internet of thing. The level of broadband penetration is negatively associated to:

- Government procurement of advanced technology products: is a measure that consider the ability of a
 government to promote technological innovation through the public demand. There is a negative
 relationship between "Broadband penetration" and the level of "Government Procurement of Advanced
 Technology Products". This negative relationship can be interpreted in the sense that the public demand
 of technological innovation is not sufficient to promote a widespread diffusion of internet among European
 countries.
- Sales impacts: is an indicator that is created as the sum of three different measures that are "Exports of medium and high-tech products", "Exports of knowledge intensive services", "Sales due to innovation activities". There is a negative relationship between Broadband penetration and sales impacts. This negative relationship can be interpreted considering that European countries, even when export, do not use digitalization and e-commerce and prefer traditional and classical methodologies.
- Firm investments: is based on three different indicators that are R&D investments, non-R&D investments, and the investment of R&D to improve the skills of human resources. There is a negative relationship between firm investment and broadband penetration. The negative relationship means that the investment of the firms does not depend on the ability of a country to promote a widespread and efficient diffusion of internet network.
- Opportunity-driven entrepreneurship: is a ratio that measures the relationship between persons that create
 enterprises to improve their income and persons that creates enterprises since they have no other
 opportunity to work. The fact that there is a negative relationship between broadband penetration and
 opportunity driven entrepreneurship means that the level of digitalization in the European Countries is not
 sufficient to promote the creation of new firms that operate for profits maximizing technological
 improvements in the digital economy.
- Most-cited publications: is ratio in which at the denominator there is the number of scientific publication in
 the top 10 most cited publication worldwide and the denominator is the total number of publication. The
 fact that there is a negative relationship between broadband penetration and most-cited publication means
 that the impact of digitalization of the efficiency of research system is not sufficient to promote an increase
 in the quality of publications at an international level.
- Rule of law: is a measure that capture the ability of economic agent to trust public institutions and their
 ability of governments to promote the quality of contract enforcement, property 4 rights, police services,
 and courts and act in the reduction of crime and violence. The negative relationship between broadband
 penetration and rule of law means that public institutions are not able to apply the digital services to justice
 and public order. Probably an improvement in the ability of government to efficiently use digital services
 to promote the role of law could have a positive effect in strengthening the of rule of law.

Sintesys of the Estimations with Dynamic Panel , Panel Data with Fixed Effects, Panel Data with Random Effects, Pooled OIS, and WLS. **Panel Data Fixed** Panel Data Random Dynamic Panel Pooled OLS Variables Effects Effects WLS Value p-value Value Label p-value Value p-value p-value Value Broadband Penetration A5 2,11448 0.829343 Costant A22 *** -1,75901 *** -1.69918 -1.87111 advanced technology products (SD) ☆-0,434044 Firm investments -0,446602 -0,615974 -0,55982 A18 0,427743 *** 0,440885 *** ** *** Enterprises providing ICT training 0.365166 *** 1,06707 *** Innovation-friendly environment 1,11725 1.11032 1,09628 1,2152 0.56413 *** *** *** 0.439624 Innovative sales share A26 0.306926 0.29238 0,468831 ... 0,509248 *** 0.532116 *** 0,536519 *** Intellectual assets Knowledge-intensive services A31 *** *** ☆ 0,501403 0,845525 0,628133 0.820915 0.698707 exports -0,531797 Most-cited publications -0,323755 ☆-0,320932 -0,356958 Opportunity-driven A39 *** *** -0,3453 ☆-0.383583 -0,485218 entrepreneurship -0,34188 *** ** -0,449353 ** -0,51152 *** ☆-0,262638 -0,423548 Rule of law (SD) * x_{10} -0.68074 *** Sales impacts A49 -0.59254 Turnover share SMEs (SD) *** 1,48295 1.03482 2 0.671793 Broadband Penetrati

Figure 1. Synthesis of the main results of the econometric estimation realized with the usage of Panel Data with Fixed Effects, Panel Data with Random Effects, Pooled OLS, Dynamic Panel Data, and WLS.

The level of broadband penetration is positively associated with:

- Enterprises providing ICT training: is ratio between the number of firms that provide training in ICT sector for their personnel and the total number of enterprises. The idea is that the greater the number of firms that train their human resources in ICT skills the greater the ability of firms to perform in the digital economy with positive effects either in the sense of profits either in the sense of GDP. The presence of a positive relationship between the broadband penetration and the number of firms that among European countries promote ICT skills for their personnel shows that the digitalization through infrastructure improve the level of human capital. Furthermore, this positive relationship can also better understood considering the role of MOOCs platforms and remote learning that are used to professional training in the digital economy.
- Innovative sales share: is a ration between the total turnover of new products and total turnover for all enterprises. The indicator shows the level of turnover associated either with new products either with the diffusion of new technologies. The existence of a positive relationship between innovative sales share and the level of broadband penetration shows that the investment in broadband penetration has effective a positive role in promoting new products and services and in the diffusion of new technologies in the markets. Innovative sales share can also be considered as a tool that shows the presence of a higher degree of customer innovativeness that can effectively sustain the marketization of new products and services.
- Intellectual assets: is a measure that capture three different forms of intellectual propriety rights that are PCT patent applications, trademark application and design application. The fact that there is a positive relationship between broadband penetration and the value of intellectual assets means that the digitalization of the economies among European Countries is positively associated to the promotion of a knowledge economy based on intangibles assets. The strengthening of the knowledge economy is either the condition either the output of the digitalization of the economy and can be efficiently realized in connection with an increase in the value of human resources at a country and international level.
- Knowledge-intensive services export: is ratio between the export of knowledge intensive services and the total export of services. The indicator can be considered as a measure of the ability of a country to export knowledge intensive services for example in the sector of transportation, finance, research and developments, ICT, media, and creative industries. The ability of a country to export knowledge intensive services is a consequence of the general degree of innovativeness of an economy. Furthermore, a higher degree of knowledge intensive services is also a signal of the ability of a country to actively participate in knowledge intensive global value chains. The positive relationship between broadband penetration and know-ledge intensive services exports means that effectively the production of high quality services in the knowledge economy requires a higher degree of digitalization.
- Turnover share SMEs: is a ratio between the turnover in SMEs and the total turnover of the business economy except financial activities. There is a positive relationship between broadband penetration and turnover share in SMEs. This means that the SMEs have greater benefits from broadband penetration in respect to big corporations. The economic opportunities for SMEs improve in the presence of higher level of broadband penetration. This result can be better understood considering the network effects and

positive externalities offered by internet and the relative low cost of the ICT in respect to other innovations that are costlier such as in the case of the pharmaceutical and automotive sectors. The digital technologies are more sustainable for SMEs, either in the sense of financial resources either in the sense of know-how. Specifically, broadband penetration can promote SMEs' competitiveness and productivity creating the conditions for better results in terms of turnover at a national level among European countries.

Innovation-friendly environment: is measure that capture the ability of firms to operate in an environment that can sustain innovation and digital firms. There is a positive relationship between broadband innovation and innovation friendly environment. The positive relationship can be in part explained considering that the presence of a digital business culture, the development of human resources that can work in ICT sectors with STEM skills, and even the possibility to have access to digital markets are effectively related to the presence of high level of broadband penetration. Furthermore, there is a recursive effect between the innovation-friendly environment and the broadband penetration: at the first stage the broadband penetration promotes a more developed innovation friendly environment and in a second stage the innovation-friendly environment expresses a demand for improvements in broadband penetration.

2. Clusterization

The clusterization model is realized with the application of the k-Means algorithm optimized with the silhouette coefficient. We found three different clusters that are:

- Cluster 1: Slovakia, United Kingdom, Norway, Bulgaria, France, Germany, Czechia, Bosnia and Herzegovina, Hungary, Austria, Poland, Estonia, Montenegro, Ukraine, North Macedonia, Romania, Ireland, Slovenia, Malta, Turkey;
- Cluster 2: Lithuania, Finland, Netherlands, Belgium, Sweden, Portugal, Denmark, Luxembourg, Switzerland, Latvia, Iceland, Spain;
- Cluster 3: Greece, Serbia, Croatia, Cyprus, Italy.

We can create an order of clusters based on the value of the median. Specifically, the value of the median of "*Broadband Penetration*" for the cluster 2 is equal to 197,48, the value of the median for 6 the cluster 1 is equal to 132,08, and the value of the median for the cluster 3 is equal to 118,25. Based on the value of the median we can build the same order: C2>C1>C3.

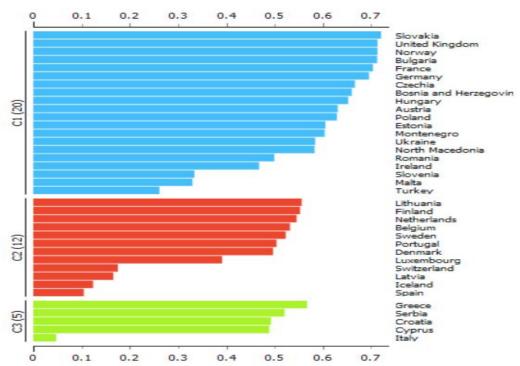


Figure 2. The Silhouette Coefficient of the Three Clusters Analyzed

3. Machine Learning and Predictions

We use eight different typologies of machine learning algorithms to choose the best predictor of the level of "Broadband Penetration". We evaluate the algorithms based on their ability to maximize R-squared and to minimize "Mean Absolute Error", "Mean Squared Error", "Root Mean Squared Error", "Mean Signed Difference". To test the algorithms, we use data from the European Innovation Scoreboard for the period 2014-2021 for 37 countries. 70% of the dataset has been used for the learning rate and the remaining 30% has been used to effectively predict the degree of "Broadband Penetration" in Europe. We found obtain the following ranking of algorithms:

- Polynomial Regression with a payoff equal to 10;
- Probabilistic Neural Network-PNN with a payoff equal to 11;
- Artificial Neural Network-ANN with a payoff equal to 19;
- Gradient Boosted Tree with a payoff equal to 20;
- *Tree Ensemble* with a payoff equal to 24;
- Linear Regression with a payoff equal to 27;
- Simple Regression Tree with a payoff equal to 32;
- Random Forest with a payoff equal to 37.

Figure 3. Ranking of algorithms in terms of performance for the maximization of R-squared and the minimization of errors.

| | Ranking of Algorithms for Performance in Terms of Maximization of R^2 and Minimization of Errors | | | | | | | |
|------|--|-----|---------------------|--------------------|-------------------------|------------------------|-----|--|
| Rank | Algorithms | R^2 | Mean Absolute Error | Mean Squared Error | Root Mean Squared Error | Mean Signed Difference | Sum | |
| 1 | Polynomial Regression | 1 | 1 | 1 | 1 | 6 | 10 | |
| 2 | PNN | 2 | 3 | 2 | 2 | 2 | 11 | |
| 3 | ANN | 4 | 2 | 3 | 3 | 7 | 19 | |
| 4 | Gradient Boosted Tree | 3 | 4 | 5 | 5 | 3 | 20 | |
| 5 | Tree Ensemble | 5 | 6 | 6 | 6 | 1 | 24 | |
| 6 | Linear Regression | 6 | 5 | 4 | 4 | 8 | 27 | |
| 7 | Simple Regression Tree | 7 | 7 | 7 | 7 | 4 | 32 | |
| 8 | Random Forest | 8 | 8 | 8 | 8 | 5 | 37 | |

The Polynomial Regression algorithm has predicted an increase of the level of "*Broadband Penetration*" in Europe equal to 10,4% on annual bases. Specifically, the algorithm has predicted the following values of "*Broadband Penetration*":

- A reduction for Austria from 130,38 to 117,38 equivalent to -9,97%;
- A reduction for Germany from 144,91 to 139,91 equivalent to -3,45%;
- An increase for Denmark from 202,48 to 208,96 equivalent to 3,20%;
- A reduction for Spain from 197,48 to 174,11 equivalent to -11,84%;
- A reduction for Croatia from 109,54 to 78,85 equivalent to -28,02%;
- A reduction for Iceland from 173,21 to 166,92 equivalent to -3,62%;
- An increase for Italy from 114,02 to 118,79 equivalent to 4,18%;
- A reduction for North Macedonia from 104,88 to 102,19 equivalent to -2,57%;
- An increase for Poland from 154,92 to 161,9 equivalent to 4,51%;
- An increase for Ukraine for 91.1 to 139.42 equivalent to +53.03%;
- An increase for United Kingdom from 134,16 to 140,86 equivalent to +4,99%.

Generally, the level of broadband penetration is expected to growth in Europe even if an heterogenous way among different countries.

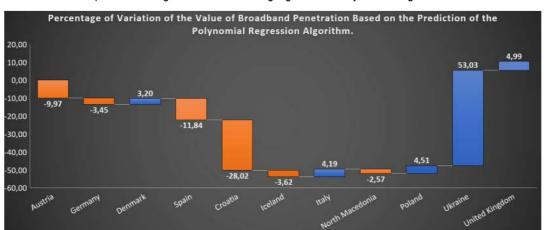


Figure 4. Percentage variation of the value of Broadband penetration among different countries in Europe as predicted using the machine learning algorithm of Polynomial Regression.

Conclusions

In this article we estimate the determinants of broadband penetration in Europe. We use data from the European Innovation Scoreboard of the European Commission for 37 countries in the period 2010-2019. In the first paragraph we present a theoretical introduction considering the role of broadband penetration as a tool to promote economic growth. Empirical studies have shown the presence of a positive relationship between broadband penetration and economic growth either in developed either in developing countries.

In particular, the effect of broadband penetration on economic growth is greater in countries with less income inequalities and a higher level of human capital and investments in intangible assets such as R&D. In the second paragraph we estimate the level of broadband penetration in European countries using different econometric models i.e. We apply Panel Data with Fixed Effects, Panel Data with Random Effects, WLS, OLS and Dynamic Panel. The results obtained are consistent among different models and statistically significant. We found that the level of "Broadband Penetration" in Europe is positively associated to "Enterprises Providing ICT Training", "Innovative Sales Share", "Intellectual Assets", "Knowledge-Intensive Service Exports", "Turnover Share SMEs", "Innovation Friendly Environment" and negatively associated with "Government procurement of advanced technology products", "Sales Impact", "Firm Investments", "Opportunity-Driven Entrepreneurship", "Most Cited Publications", "Rule of Law".

The cluster analysis with k-Means algorithm optimized with the Silhouette Coefficient shows the presence of three clusters that substantially coincide, with some exceptions, with Northern, Central and Southern Europe. Finally, we apply eight machine learning algorithms to predict the level of "*Broadband Penetration*" in Europe. We find that the Polynomial Regression algorithm is the best predictor and that the level of the broadband penetration is expected to increase of 10,4%. Our results show that if policy makers are interested in promoting the economic growth then they should invest more in broadband penetration at a European level.

References

- [1] Agboje, O.E., Adedoyin, S.O., and Ndujiuba, C.U. 2017. State of fiber optic networks for internet broadband penetration in Nigeria-a review, *International Journal of Optoelectronic Engineering*, 7(1): pp. 1-12.
- [2] Aissaoui, N. 2017. Is inequality harmful for broadband diffusion and economic growth, *Asian Economic and Financial Review*, 7(8): 799-808.
- [3] Akinwale, O.Y., Sanusi, A., and Surujlal, J. 2018. An empirical analysis of information and communication technology (ICT) and economic growth in Nigeria, *International Journal of eBusiness and eGovernment Studies*, 10(1): 129-142.
- [4] Alam, T.F., Sultana, N., and Rayhan, M.I. 2019. Structural equation modeling: an application of broadband penetration and GDP growth in Asia, *Journal of Economic Structures*, 8(1)-17. DOI: https://doi.org/10.1186/s40008-019-0148-0
- [5] Alderete, M.V. 2017. An approach to the broadband effect on Latin American growth: A structural model, *Cuadernos de Economía*, 36 (71): 549-569.

- [6] Bahia, K., Castells, P., and Pedrós, X. 2019. The impact of mobile technology on economic growth: global insights from 2000-2017 developments., *30th European Regional ITS Conference at Helsinki*, no. 205164.
- [7] Benseny, J., Töyli, J., Hämmäinen, H., and Arcia-Moret, A. 2019. The mitigating role of regulation on the concentric patterns of broadband diffusion: The case of Finland. *Telematics and Informatics*, 41: 139-155. DOI: https://doi.org/10.1016/j.tele.2019.04.008
- [8] Chattopadhyay, A., and Majumdar, M. 2019. A bottom-up hierarchical forecasting of broadband penetration in India using ARIMA & DES approaches, *International Academic Journal of Business Management*, 6(1): 110-119. DOI: https://doi.org/10.9756/IAJBM/V6I1/1910014
- [9] Costantiello, A., and Leogrande, A. 2020. The innovation-employment nexus in Europe., *American Journal of Humanities and Social Sciences Research (AJHSSR)*, 4(11): 166-187.
- [10] Costantiello, A., Laureti, L. De Cristoforo, G. and Leogrande, A. 2021. *The innovation-sales growth nexus in Europe*, University Library of Munich, Germany. DOI: http://dx.doi.org/10.2139/ssrn.3933407
- [11] David, O.O., and Grobler, W. 2020. Information and communication technology penetration level as an impetus for economic growth and development in Africa, *Economic Research-Ekonomska istraživanja*, 33(1): 1394-1418.
- [12] Djunaedi, A.Z. 2021. Digitalization impact on growth & human capital: Indonesia broadband plan case study, *Review of Business, Accounting, & Finance*, 1(3): 299-309.
- [13] Dwiardi, A.R. 2020. Analysis of the needs of ICT Ecosystems to Support the Acceleration of Internet Fixed Broadband Penetration. *Jurnal Penelitian Pos dan Informatika*, 10(1): 41-58.
- [14] Geraci, A., Nardotto, M., Reggiani, T., and Sabatini, F. 2018. *Broadband Internet and social capital*, arXiv preprint arXiv:1810.04575
- [15] Habibi, F., Zabardast, M.A. 2020. Digitalization, education and economic growth: A comparative analysis of Middle East and OECD countries, *Technology in Society*, 63. DOI: https://doi.org/10.1016/j.techsoc.2020.101370
- [16] Hendriks Vettehen, P., Troost, J., Boerboom, L., Steijaert, M., and Scheepers, P. 2020. The relationship between media content preferences and political participation in 25 European countries: The moderating role of broadband penetration and broadband access, *Communication Research*, 47(7): 967-987. DOI: https://doi.org/10.1177/0093650217701033
- [17] Imasheva, I.Y., and Kramin, T. 2020. Impact of broadband internet on the economic growth of the Russian regions. in first international Volga region conference on economics, *Humanities and Sports*, 19(5): 26-28. DOI: https://doi.org/10.2991/aebmr.k.200114.006
- [18] Jung, J., and López-Bazo, E. 2020. On the regional impact of broadband on productivity: The case of Brazil, *Telecommunications Policy*, 1(101826): 44. DOI: https://doi.org/10.1016/j.telpol.2019.05.002
- [19] Kabaklarli, E., and Atasoy, B.S. 2019. Broadband infrastructure and economic growth: a panel data approach for selected countries, DIEM: *Dubrovnik International Economic Meeting*, 4(1): 105-114.
- [20] Kurniawati, M.A. 2021. Analysis of the impact of information communication technology on economic growth: empirical evidence from Asian countries, *Journal of Asian Business and Economic Studies*, DOI: https://doi.org/10.1108/JABES-07-2020-0082
- [21] Lange, M.R. 2017. Tariff diversity and competition policy: Drivers for broadband adoption in the European Union, *Journal of Regulatory Economics*, 52(3): 285-312. DOI: https://doi.org/10.1007/s11149-017-9344-8
- [22] Leogrande A., and Costantiello, A. 2021. Human resources in Europe. Estimation, clusterization, machine learning and prediction. machine learning and prediction. *American Journal of Humanities and Social Sciences Research (AJHSSR)*, 5(9): 240-259.
- [23] Leogrande, A., Massaro, A., and Galiano, A.M. 2020. The attractiveness of European research systems, *American Journal of Humanities and Social Sciences Research (AJHSSR)*, 4 10): 72-101.

- [24] Leogrande, A., Massaro, A., and Galiano, A.M. 2020. The impact of R&D investments on corporate performance in European Countries, *American Journal of Humanities and Social Sciences Research* (AJHSSR), 4(7): 186-201.
- [25] Maneejuk, P. Yamaka, W. 2020. An analysis of the impacts of telecommunications technology and innovation on economic growth, *Telecommunications Policy*, 44(10). DOI: https://doi.org/10.1016/j.telpol.2020.102038
- [26] Na, H.S., Hwang, J., and Kim, H. 2020. Digital content as a fast Internet diffusion factor: Focusing on the fixed broadband Internet, *Information Development*, 36(1): 97-111. DOI: https://doi.org/10.1177/0266666918811878
- [27] Oztunc, H., Demirbas, E., and Orhan, M.A. 2019. Comparative study on the impact of telecommunication investment on economic growth in EU Vis-à-vis Non-EU OECD countries: A dynamic panel data analysis, *Asian Journal of Economics, Business and Accounting*, 1-16. DOI: https://doi.org/10.9734/ajeba/2019/v12i330152
- [28] Romer, P.M. 1994. The origins of endogenous growth, *Journal of Economic Perspectives*, 8(1): 3-22. DOI: https://doi.org/10.1257/jep.8.1.3
- [29] Schumpeter, J.A. 1934. The theory of economic development: an inquiry into profits, capital, credit, interest, and the business cycle, New Brunswick, New Jersey: Transaction Books.
- [30] Shin, S., and Koh, J. 2017. Analysis of mobile broadband service penetration in South Korea, *Journal of Computer Information Systems*, 57(1): 31-38, DOI: https://doi.org/10.1080/08874417.2016.1181491
- [31] Shodiev, T., Turayey, B., and Shodiyev, K. 2021. ICT and economic growth nexus: Case of central Asian countries, *Procedia of Social Sciences and Humanities*, 1. DOI: https://doi.org/10.21070/pssh.v1i.37
- [32] Solow, R.M. 1956. A contribution to the theory of economic growth, *The Quarterly Journal of Economics*, 70(1): 65-94. DOI: https://doi.org/10.2307/1884513

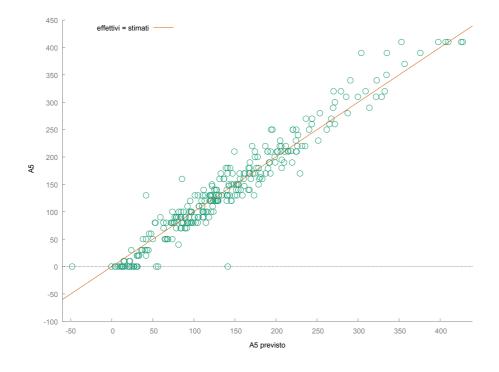
APPENDIX

Econometric Estimation

| Panel dinamico a un passo, usando 288 osservazioni | | |
|--|--|--|
| Incluse 36 unità cross section | | |
| Matrice H conforme ad Ox/DPD | | |
| Variabile dipendente: A5 | | |

| - | Coefficiente | Errore Std. | Z | p-value |
|----------------------|--------------|-------------|-------------------|-------------|
| A5(-1) | -0,286854 | 0,131576 | -2,180 | 0,0292 ** |
| const | 14,3267 | 4,39060 | 3,263 | 0,0011 *** |
| A22 | -1,69918 | 0,167796 | -10,13 | <0,0001 *** |
| A18 | -0,434044 | 0,0939110 | -4,622 | <0,0001 *** |
| A15 | 0,179466 | 0,0737442 | 2,434 | 0,0149 ** |
| A25 | 1,11725 | 0,205615 | 5,434 | <0,0001 *** |
| A26 | 0,306926 | 0,0746325 | 4,112 | <0,0001 *** |
| A29 | 0,432916 | 0,112743 | 3,840 | 0,0001 *** |
| A31 | 0,501403 | 0,177975 | 2,817 | 0,0048 *** |
| A36 | -0,320932 | 0,112613 | -2,850 | 0,0044 *** |
| A39 | -0,383583 | 0,142815 | -2,686 | 0,0072 *** |
| A48 | -0,262638 | 0,153474 | -1,711 | 0,0870 * |
| A49 | -0,592540 | 0,172724 | -3,431 | 0,0006 *** |
| A58 | 0,671793 | 0,191483 | 3,508 | 0,0005 *** |
| Somma quadr. Residui | 156640,1 | E.S. | della regressione | 23,90981 |

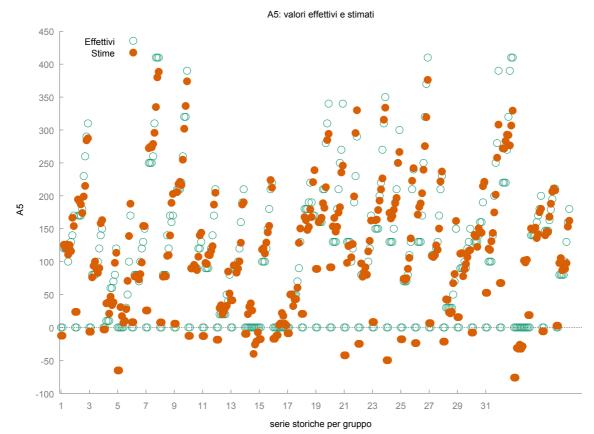
| Numero di strumenti = 25 |
|--|
| Test per errori AR(1): z = 1,57337 [0,1156] |
| Test per errori AR(2): z = 1,50915 [0,1313] |
| Test di sovra-identificazione di Sargan: Chi-quadro(11) = 63,5395 [0,0000] |
| Test (congiunto) di Wald: Chi-quadro(13) = 879,972 [0,0000] |



| Effetti fissi, usando 360 osservazioni | | |
|--|--|--|
| Incluse 36 unità cross section | | |
| Lunghezza serie storiche = 10 | | |
| Variabile dipendente: A5 | | |

| | Coefficiente | Errore Std. | rapporto t | p-value |
|-----------------------|--------------|-------------|------------------------|-------------|
| const | 0,829343 | 3,57650 | 0,2319 | 0,8168 |
| A22 | -1,75901 | 0,172232 | -10,21 | <0,0001 *** |
| A18 | -0,400542 | 0,0950476 | -4,214 | <0,0001 *** |
| A15 | 0,418365 | 0,0584097 | 7,163 | <0,0001 *** |
| A25 | 1,11032 | 0,0688436 | 16,13 | <0,0001 *** |
| A26 | 0,292380 | 0,0905978 | 3,227 | 0,0014 *** |
| A29 | 0,509248 | 0,114761 | 4,437 | <0,0001 *** |
| A31 | 0,845525 | 0,140959 | 5,998 | <0,0001 *** |
| A36 | -0,323755 | 0,147503 | -2,195 | 0,0289 ** |
| A39 | -0,485218 | 0,0706066 | -6,872 | <0,0001 *** |
| A48 | -0,423548 | 0,208933 | -2,027 | 0,0435 ** |
| A49 | -0,884347 | 0,181574 | -4,870 | <0,0001 *** |
| A58 | 1,36469 | 0,126128 | 10,82 | <0,0001 *** |
| Media var. dipendente | 112,0833 | | SQM var. dipendente | 101,5265 |
| Somma quadr. Residui | 314740,0 | | E.S. della regressione | 31,76133 |
| R-quadro LSDV | 0,914945 | | R-quadro intra-gruppi | 0,858641 |
| LSDV F(47, 312) | 71,40907 | | P-value(F) | 3,1e-140 |
| Log-verosimiglianza | -1730,030 | | Criterio di Akaike | 3556,059 |
| Criterio di Schwarz | 3742,592 | | Hannan-Quinn | 3630,228 |
| Rho | 0,634619 | | Durbin-Watson | 0,627291 |

| Fest congiunto sui regressori - | | |
|--|--|--|
| Statistica test: F(12, 312) = 157,929 | | |
| con p-value = P(F(12, 312) > 157,929) = 1,12046e-124 | | |
| Fest per la differenza delle intercette di gruppo - | | |
| Ipotesi nulla: i gruppi hanno un'intercetta comune | | |
| Statistica test: F(35, 312) = 10,4881 | | |
| con p-value = $P(F(35, 312) > 10,4881) = 3,82213e-035$ | | |



| Effetti casuali (GLS), usando 360 osservazioni | | |
|--|--|--|
| Incluse 36 unità cross section | | |
| Lunghezza serie storiche = 10 | | |
| Variabile dipendente: A5 | | |

| | Coefficiente | Errore Std. | Z | p-value | |
|-----------------------|--------------|-------------|------------------------|---------|----------|
| const | 1,07776 | 6,67624 | 0,1614 | 0,8718 | |
| A22 | -1,78254 | 0,152514 | -11,69 | <0,0001 | *** |
| A18 | -0,446602 | 0,0863458 | -5,172 | <0,0001 | *** |
| A15 | 0,427743 | 0,0525809 | 8,135 | <0,0001 | *** |
| A25 | 1,09628 | 0,0639990 | 17,13 | <0,0001 | *** |
| A26 | 0,338208 | 0,0848260 | 3,987 | <0,0001 | *** |
| A29 | 0,532116 | 0,104139 | 5,110 | <0,0001 | *** |
| A31 | 0,820915 | 0,127135 | 6,457 | <0,0001 | *** |
| A36 | -0,356958 | 0,136230 | -2,620 | 0,0088 | *** |
| A39 | -0,440800 | 0,0660597 | -6,673 | <0,0001 | *** |
| A48 | -0,449353 | 0,186766 | -2,406 | 0,0161 | ** |
| A49 | -0,884979 | 0,162912 | -5,432 | <0,0001 | *** |
| A58 | 1,38740 | 0,124454 | 11,15 | <0,0001 | *** |
| Media var. dipendente | 112,0833 | | SQM var. dipendente | | 101,5265 |
| Somma quadr. residui | 744088,8 | | E.S. della regressione | | 46,24053 |
| Log-verosimiglianza | -1884,904 | | Criterio di Akaike | _ | 3795,808 |
| Criterio di Schwarz | 3846,327 | | Hannan-Quinn | _ | 3815,895 |
| rho | 0,634619 | | Durbin-Watson | | 0,627291 |

Varianza 'between' = 1150,84

Varianza 'within' = 1008,78

Theta usato per la trasformazione = 0,716113

Test congiunto sui regressori
Statistica test asintotica: Chi-quadro(12) = 1975,55

con p-value = 0

Test Breusch-Pagan
Ipotesi nulla: varianza dell'errore specifico all'unità = 0

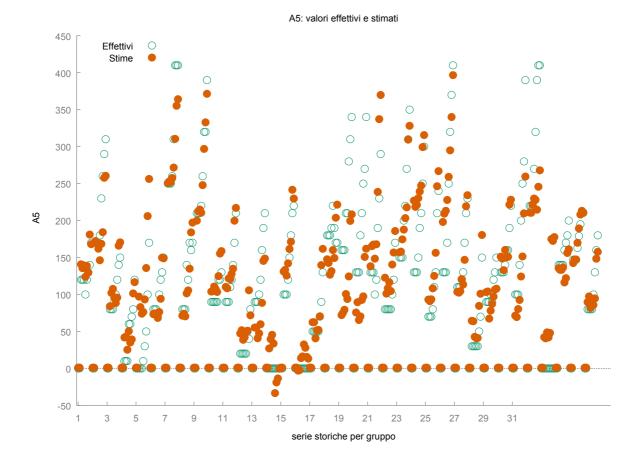
Statistica test asintotica: Chi-quadro(1) = 299,802

con p-value = 3,63909e-067

Test di Hausman
Ipotesi nulla: le stime GLS sono consistenti

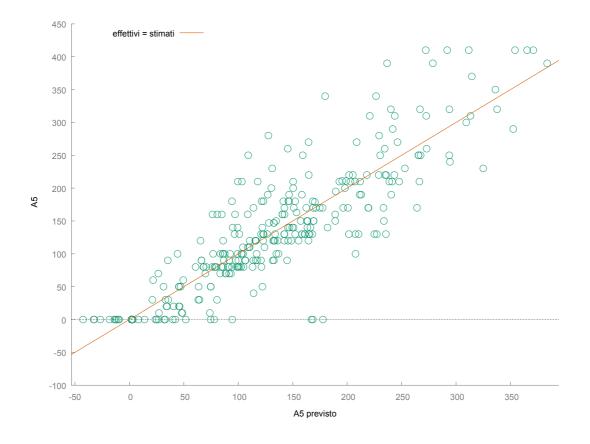
Statistica test asintotica: Chi-quadro(12) = 13,2734

con p-value = 0,349489



| Pooled OLS, usando 360 osservazioni | | |
|-------------------------------------|--|--|
| Incluse 36 unità cross section | | |
| Lunghezza serie storiche = 10 | | |
| Variabile dipendente: A5 | | |

| | Coefficiente | Errore Std. | rapporto t | p-value | |
|-----------------------|--------------|--------------------------|------------------------|------------|----------|
| const | 2,11448 | 4,59343 | 0,4603 | 0,6456 | |
| A22 | -1,87111 | 0,116558 | -16,05 | <0,0001 | *** |
| A18 | -0,615974 | 0,0710567 | -8,669 | <0,0001 | *** |
| A15 | 0,440885 | 0,0418835 | 10,53 | <0,0001 | *** |
| A25 | 1,06707 | 0,0573275 | 18,61 | <0,0001 | *** |
| A26 | 0,564130 | 0,0799626 | 7,055 | <0,0001 | *** |
| A29 | 0,536519 | 0,0853741 | 6,284 | <0,0001 | *** |
| A31 | 0,698707 | 0,102075 | 6,845 | <0,0001 | *** |
| A36 | -0,531797 | 0,118725 | -4,479 | <0,0001 | *** |
| A39 | -0,240413 | 0,0602629 | -3,989 | <0,0001 | *** |
| A48 | -0,511520 | 0,145048 | -3,527 | 0,0005 | *** |
| A49 | -0,820701 | 0,128934 | -6,365 | <0,0001 | *** |
| A58 | 1,48295 | 0,162968 | 9,100 | <0,0001 | *** |
| Media var. dipendente | 112,0833 | | SQM var. dipendente | 1 | 01,5265 |
| Somma quadr. residui | 685047,7 | | E.S. della regressione | e 44,43198 | |
| R-quadro | 0,814874 | 814874 R-quadro corretto | | 0,808472 | |
| F(12, 347) | 127,2833 | | P-value(F) | 4 | 1,2e-119 |
| Log-verosimiglianza | -1870,023 | | Criterio di Akaike | 3 | 766,046 |
| Criterio di Schwarz | 3816,565 | | Hannan-Quinn | 3 | 786,133 |
| rho | 0,912159 | | Durbin-Watson | 0 | ,339898 |

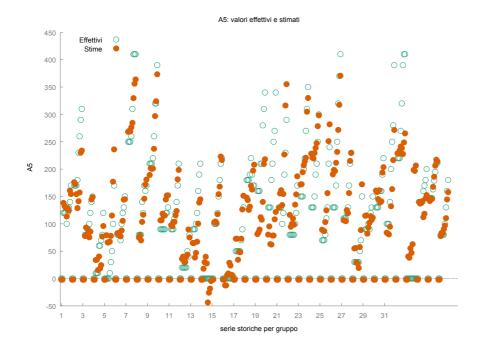


45,99399

| WLS, usando 360 osservazioni | |
|---|--|
| Incluse 36 unità cross section | |
| Variabile dipendente: A5 | |
| Pesi basati sulle varianze degli errori per unità | |
| • | |

| | Coefficiente Errore Std. | | rapporto t | p-value | | |
|-----------------------|-----------------------------|-----------|-------------|--------------------|---------|----------|
| const | -2,14400 | 2,07 | 945 | -1,031 | 0,3032 | |
| A22 | -1,95799 | 0,062 | 4419 | -31,36 | <0,0001 | *** |
| A18 | -0,559823 | 0,031 | 4576 | -17,80 | <0,0001 | *** |
| A15 | 0,365166 | 0,020 | 4196 | 17,88 | <0,0001 | *** |
| A25 | 1,21520 | 0,032 | 9852 | 36,84 | <0,0001 | *** |
| A26 | 0,439624 | 0,035 | 7728 | 12,29 | <0,0001 | *** |
| A29 | 0,468831 | 0,046 | 1429 | 10,16 | <0,0001 | *** |
| A31 | 0,628133 | 0,059 | 0076 | 10,64 | <0,0001 | *** |
| A36 | -0,406173 | 0,063 | 5971 | -6,387 | <0,0001 | *** |
| A39 | -0,345296 | 0,038 | 5359 | -8,960 | <0,0001 | *** |
| A48 | -0,341875 | 0,060 | 0756 | -5,691 | <0,0001 | *** |
| A49 | -0,680742 | 0,059 | 8350 | -11,38 | <0,0001 | *** |
| A58 | 1,03482 | 0,077 | 0871 | 13,42 | <0,0001 | *** |
| | Statistich | ne basate | sui dati po | onderati: | | |
| Somma quadr. residui | 28 | 0,6645 | E.S. de | ella regressione | | 0,899351 |
| R-quadro | -quadro 0,94 | | | dro corretto | | 0,946935 |
| F(12, 347) | 53 | 4,8533 | P-valu | e(F) | | 1,7e-215 |
| Log-verosimiglianza | -46 | -466,0079 | | Criterio di Akaike | | 958,0159 |
| Criterio di Schwarz | riterio di Schwarz 1008,535 | | | n-Quinn | | 978,1033 |
| | Statistic | he basate | sui dati o | riginali: | | |
| Media var. dipendente | 11. | 2,0833 | SQM v | ar. dipendente | | 101,5265 |

E.S. della regressione

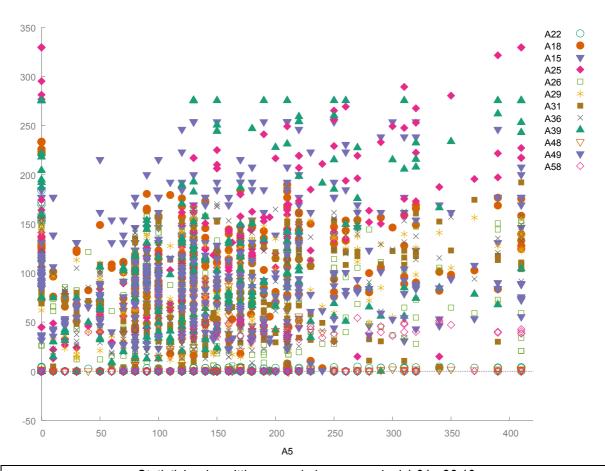


734060,3

Somma quadr. residui

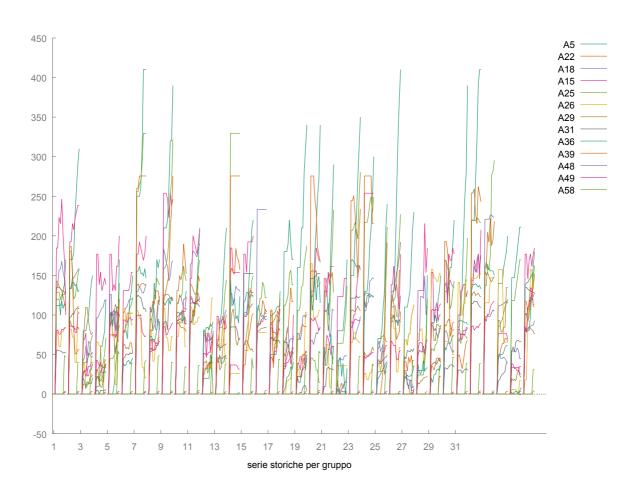


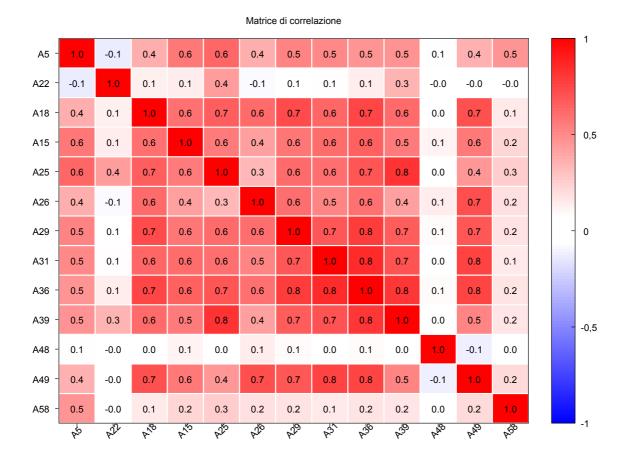
| Α | : 4-11 | | | : | | | | | | | | |
|--|--------------|-----------|---------|--------|---------|--------|---------|--|--|--|--|--|
| Analisi delle componenti principali | | | | | | | | | | | | |
| n = 360 | | | | | | | | | | | | |
| Analisi degli autovalori della matrice di correlazione | | | | | | | | | | | | |
| | | | | | | | | | | | | |
| Compo | onente Au | tovalore | | | umulata | 1 | | | | | | |
| 1 | -1 | | | | | | | | | | | |
| 2 | 1,4193 | 0,1092 | 2 0,6 | 5122 | | | | | | | | |
| 3 | 1,2267 | 0,094 | 4 0,7 | 7066 | | | | | | | | |
| 4 | 1,0484 | 0,080 | 6 0,7 | 7872 | | | | | | | | |
| 5 | 0,6916 | 0,0532 | 2 0,8 | 3404 | | | | | | | | |
| 6 | 0,4933 | 0,0379 | 9 0,8 | 3784 | | | | | | | | |
| 7 | 0,4457 | 0,0343 | 3 0,9 | 9127 | | | | | | | | |
| 8 | 0,3546 | 0,027 | 3 0,9 | 9399 | | | | | | | | |
| 9 | 0,2725 | 0,0210 | | 9609 | | | | | | | | |
| 10 | 0,2296 | 0,017 | | 9786 | | | | | | | | |
| 11 | 0,1274 | 0,0098 | | 9884 | | | | | | | | |
| 12 | 0,0984 | 0,007 | | 9959 | | | | | | | | |
| 13 | 0,0528 | 0,004 | | 0000 | | | | | | | | |
| | -,0020 | 2,301 | .,. | • | | | | | | | | |
| Autove | ettori (pesi | della cor | nnonent | te) | | | | | | | | |
| , | cton (posi | aona 601 | | , | | | | | | | | |
| | PC1 | PC2 | PC3 | PC4 | PC5 | PC6 | PC7 | | | | | |
| A5 | -0,252 | -0,175 | | -0,083 | -0,384 | | | | | | | |
| A22 | -0,232 | 0,746 | 0,004 | 0,014 | 0,402 | 0,189 | | | | | | |
| | | | | | | | | | | | | |
| A18 | -0,325 | -0,007 | -0,188 | 0,022 | 0,178 | 0,025 | | | | | | |
| A15 | -0,294 | 0,006 | 0,123 | 0,069 | -0,195 | 0,850 | | | | | | |
| A25 | -0,313 | 0,348 | 0,213 | -0,008 | -0,114 | -0,233 | • | | | | | |
| A26 | -0,267 | | -0,199 | 0,087 | 0,439 | -0,097 | | | | | | |
| A29 | -0,334 | | -0,091 | 0,052 | -0,030 | -0,210 | | | | | | |
| A31 | | -0,032 | | | | 0,010 | • | | | | | |
| A36 | | 0,029 | | | | | | | | | | |
| A39 | | | | | | | 5 0,173 | | | | | |
| A48 | -0,028 | -0,101 | 0,208 | 0,925 | 0,140 | -0,011 | | | | | | |
| A49 | | | | -0,203 | 0,191 | 0,147 | | | | | | |
| A58 | -0,115 | -0,166 | 0,657 | -0,276 | 0,548 | -0,010 | 0 0,281 | | | | | |
| | | | | | | | | | | | | |
| | PC8 | PC9 | PC10 | PC11 | PC12 | PC13 | 3 | | | | | |
| A5 | 0,321 | 0,041 | -0,264 | 0,081 | 0,040 | -0,464 | | | | | | |
| A22 | 0,264 | 0,219 | -0,237 | 0,101 | 0,022 | -0,239 | 9 | | | | | |
| A18 | -0,580 | -0,345 | -0,163 | 0,330 | -0,155 | -0,248 | 8 | | | | | |
| A15 | -0,008 | 0,035 | 0,299 | -0,011 | -0,039 | 0,177 | 7 | | | | | |
| A25 | -0,032 | -0,296 | -0,194 | -0,342 | 0,164 | 0,579 | 9 | | | | | |
| A26 | 0,589 | -0,097 | 0,184 | 0,208 | 0,028 | 0,203 | | | | | | |
| A29 | -0,240 | 0,840 | 0,007 | -0,024 | -0,088 | 0,157 | | | | | | |
| A31 | 0,137 | | -0,415 | 0,415 | -0,282 | 0,258 | | | | | | |
| A36 | | -0,020 | 0,189 | 0,114 | 0,809 | -0,185 | | | | | | |
| A39 | | -0,140 | 0,621 | | -0,436 | -0,229 | | | | | | |
| A48 | -0,052 | | | -0,030 | -0,450 | | | | | | | |
| A49 | 0,052 | | -0,124 | | -0,052 | | | | | | | |
| | | | | | | | | | | | | |
| A58 | -0,221 | 0,025 | 0,099 | บ,บวง | -0,019 | 0,089 | | | | | | |



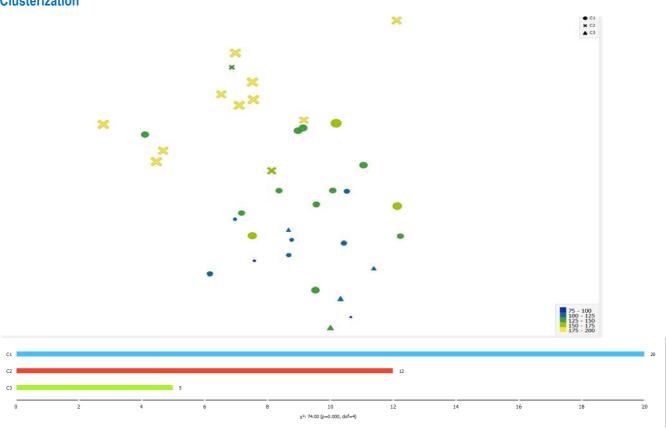
| ; | Statistiche de | escrittive, usando l | e osservazioni 1:01 - 3 | 36:10 |
|-----------|----------------|----------------------|-------------------------|----------|
| Variabile | Media | Mediana | Minimo | Massimo |
| A5 | 112,08 | 100,00 | 0,00000 | 410,00 |
| A22 | 4,3792 | 0,00000 | 0,00000 | 183,97 |
| A18 | 84,483 | 89,888 | 0,00000 | 233,51 |
| A15 | 96,325 | 96,154 | 0,00000 | 253,85 |
| A25 | 102,02 | 98,640 | 0,00000 | 329,62 |
| A29 | 65,865 | 70,699 | 0,00000 | 156,33 |
| A31 | 60,206 | 55,536 | 0,00000 | 192,21 |
| A36 | 65,732 | 58,826 | 0,00000 | 170,01 |
| A39 | 85,070 | 67,137 | 0,00000 | 275,59 |
| A48 | 6,1413 | 0,00000 | -0,76374 | 92,715 |
| A49 | 52,107 | 54,227 | -0,25937 | 134,39 |
| A58 | 6,4516 | 0,00000 | 0,00000 | 54,230 |
| Variabile | SQM | Coeff. di variazione | Asimmetria | Curtosi |
| A5 | 101,53 | 0,90581 | 0,79972 | 0,24640 |
| A22 | 24,031 | 5,4875 | 6,4562 | 40,202 |
| A18 | 61,263 | 0,72515 | 0,24491 | -0,41895 |
| A15 | 78,186 | 0,81169 | 0,19383 | -1,1612 |
| A25 | 84,060 | 0,82400 | 0,76051 | 0,18165 |
| A29 | 48,645 | 0,73855 | 0,067354 | -1,1580 |
| A31 | 48,403 | 0,80395 | 0,29595 | -1,0101 |
| A36 | 53,789 | 0,81831 | 0,22152 | -1,3515 |

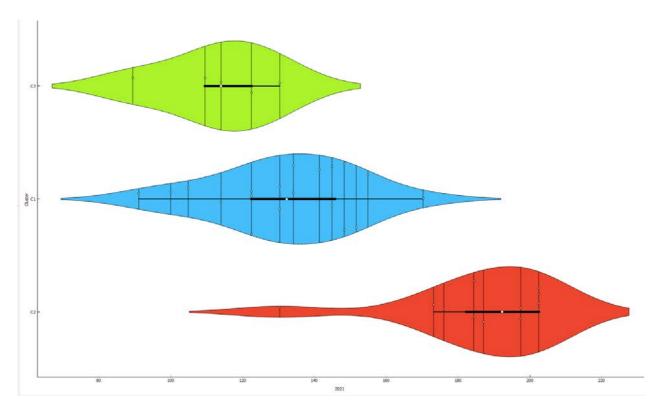
| 86,220 | 1,0135 | 0,88117 | -0,28263 |
|----------|---|--|---|
| 18,346 | 2,9873 | 3,2151 | 9,3353 |
| 41,619 | 0,79871 | -0,029395 | -1,4204 |
| 15,399 | 2,3869 | 2,0294 | 2,2966 |
| 5% Perc. | 95% Perc. | Range interquartile | Osservazioni mancanti |
| 0,00000 | 310,00 | 170,00 | 0 |
| 0,00000 | 4,0707 | 0,00000 | 0 |
| 0,00000 | 184,03 | 85,907 | 0 |
| 0,00000 | 238,08 | 161,54 | 0 |
| 0,00000 | 269,24 | 113,18 | 0 |
| 0,00000 | 145,41 | 82,152 | 0 |
| 0,00000 | 146,72 | 96,837 | 0 |
| 0,00000 | 155,39 | 98,825 | 0 |
| 0,00000 | 275,59 | 130,58 | 0 |
| 0,00000 | 56,015 | 0,85821 | 0 |
| 0,00000 | 113,43 | 88,200 | 0 |
| 0,00000 | 45,894 | 0,00000 | 0 |
| | 18,346 41,619 15,399 5% Perc. 0,00000 0,00000 0,00000 0,00000 0,00000 0,00000 0,00000 0,00000 0,00000 0,00000 0,00000 | 18,346 2,9873 41,619 0,79871 15,399 2,3869 5% Perc. 95% Perc. 0,00000 310,00 0,00000 4,0707 0,00000 184,03 0,00000 238,08 0,00000 269,24 0,00000 145,41 0,00000 146,72 0,00000 155,39 0,00000 275,59 0,00000 56,015 0,00000 113,43 | 18,346 2,9873 3,2151 41,619 0,79871 -0,029395 15,399 2,3869 2,0294 5% Perc. 95% Perc. Range interquartile 0,00000 310,00 170,00 0,00000 4,0707 0,00000 0,00000 184,03 85,907 0,00000 238,08 161,54 0,00000 269,24 113,18 0,00000 145,41 82,152 0,00000 146,72 96,837 0,00000 155,39 98,825 0,00000 275,59 130,58 0,00000 56,015 0,85821 0,00000 113,43 88,200 |

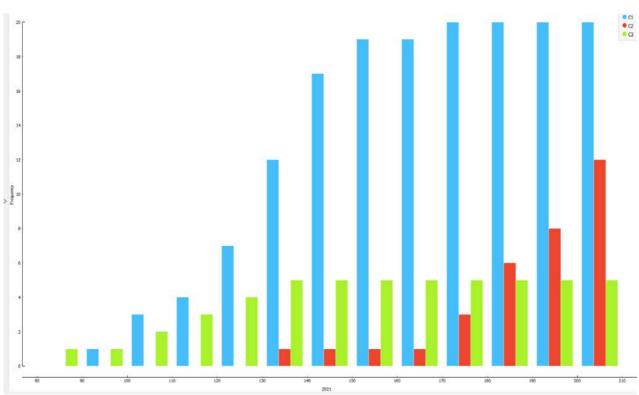


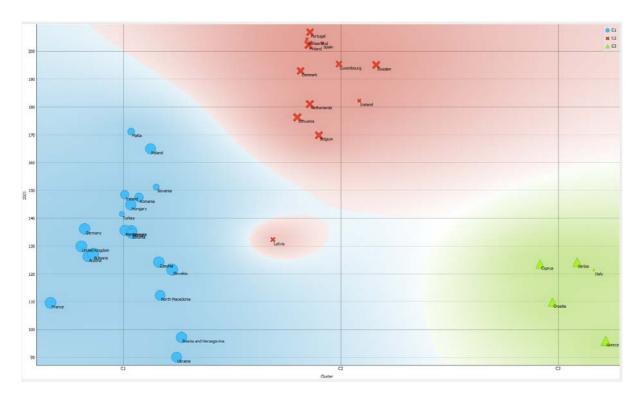


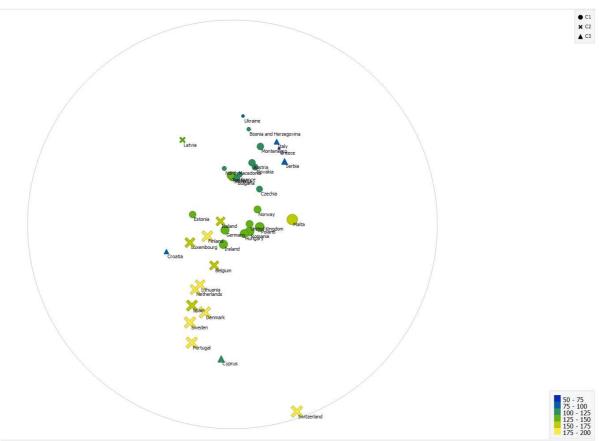
Clusterization

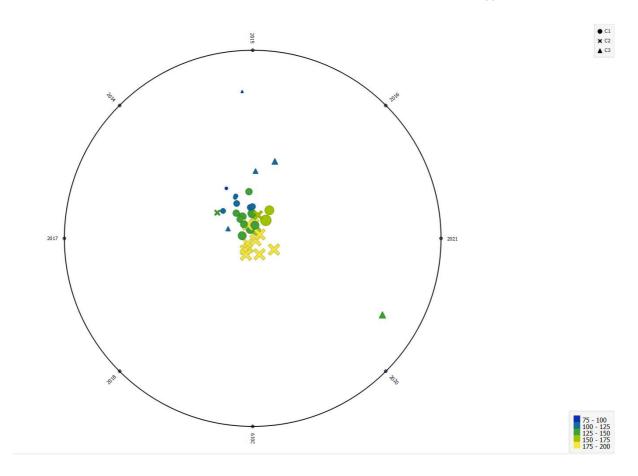


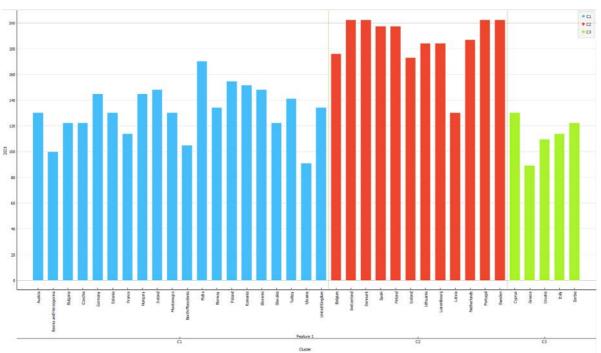


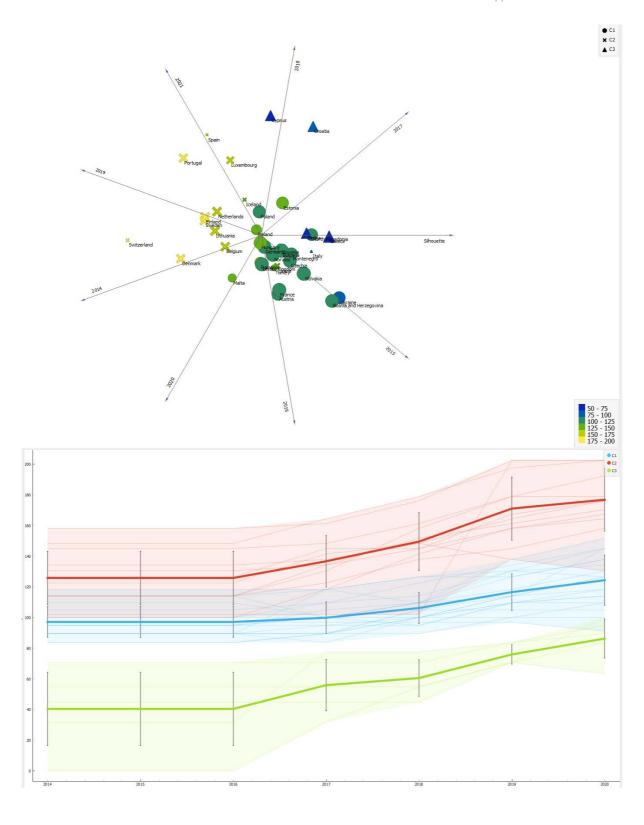




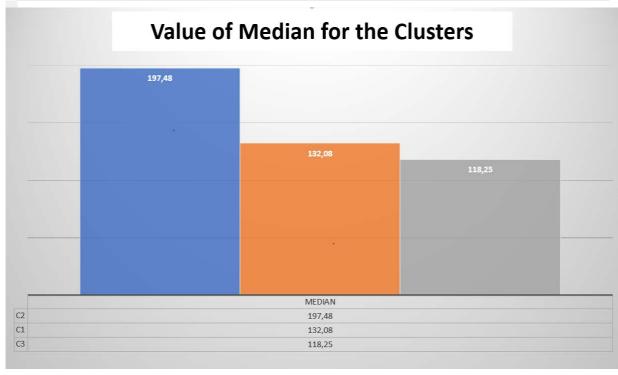








| | 2021 | Feature 1 | Cluster | Silhouette | 2014 | 2015 | 2016 | 2017 | 2018 | 2019 | 2020 |
|----|---------|-----------------|---------|------------|---------|---------|---------|---------|---------|---------|---------|
| 37 | 134.164 | United Kingdom | C1 | 0.701575 | 89.4427 | 89.4427 | 89.4427 | 89.4427 | 100 | 114.018 | 126,491 |
| 36 | 91.1043 | Ukraine | C1 | 0.680665 | 96.9536 | 96.9536 | 96.9536 | 96.9536 | 96.9536 | 96.9536 | 91.1043 |
| 35 | 141.421 | Turkey | C1 | 0.568808 | 118.322 | 118.322 | 118.322 | 118.322 | 126.491 | 130.384 | 134.164 |
| 33 | 148.324 | Slovenia | C1 | 0.59357 | 114.018 | 114.018 | 114.018 | 118.322 | 126.491 | 126.491 | 137.84 |
| 34 | 122.474 | Slovakia | C1 | 0.704062 | 94.8683 | 94.8683 | 94.8683 | 100 | 94.8683 | 109.545 | 114.018 |
| 30 | 151.658 | Romania | C1 | 0.651 | 104.881 | 104.881 | 104.881 | 109.545 | 114.018 | 130.384 | 144.914 |
| 28 | 154.919 | Poland | C1 | 0.68278 | 83.666 | 83.666 | 83.666 | 89.4427 | 104.881 | 114.018 | 144.914 |
| 27 | 134.164 | Norway | C1 | 0.701575 | 89.4427 | 89.4427 | 89.4427 | 89.4427 | 100 | 114.018 | 126.491 |
| 24 | 104.881 | North Macedonia | C1 | 0.67468 | 83.666 | 83.666 | 83.666 | 89.4427 | 104.881 | 104.881 | 104.881 |
| 23 | 130.384 | Montenegro | C1 | 0.680726 | 89.4427 | 89.4427 | 89.4427 | 89.4427 | 89.4427 | 100 | 109.545 |
| 25 | 170.294 | Malta | C1 | 0.607852 | 114.018 | 114.018 | 114.018 | 100 | 109.545 | 137.84 | 151.658 |
| 17 | 148.324 | Ireland | C1 | 0.644739 | 100 | 100 | 100 | 109.545 | 122.474 | 134.164 | 144.914 |
| 16 | 144.914 | Hungary | C1 | 0.688867 | 94.8683 | 94.8683 | 94.8683 | 100 | 109.545 | 126.491 | 137.84 |
| 8 | 144.914 | Germany | C1 | 0.696346 | 94.8683 | 94.8683 | 94.8683 | 104.881 | 109.545 | 118.322 | 130.384 |
| 14 | 114.018 | France | C1 | 0.70175 | 94.8683 | 94.8683 | 94.8683 | 94.8683 | 94.8683 | 109.545 | 109.545 |
| 10 | 130.384 | Estonia | C1 | 0.674277 | 89.4427 | 89.4427 | 89.4427 | 118.322 | 109.545 | 130.384 | 126.491 |
| 7 | 122.474 | Czechia | C1 | 0.691607 | 89.4427 | 89.4427 | 89.4427 | 83.666 | 100 | 109.545 | 114.018 |
| 4 | 122.474 | Bulgaria | C1 | 0.702373 | 89.4427 | 89.4427 | 89.4427 | 94.8683 | 100 | 109.545 | 118.322 |
| 2 | 100 | Bosnia and Her | C1 | 0.693818 | 100 | 100 | 100 | 100 | 100 | 100 | 100 |
| 1 | 130.384 | Austria | C1 | 0.679029 | 109.545 | 109.545 | 109.545 | 100 | 109.545 | 114.018 | 118.322 |
| 5 | 202.485 | Switzerland | C2 | 0.530825 | 101.98 | 101.98 | 101.98 | 101.98 | 101.98 | 202.485 | 202.485 |
| 32 | 202.485 | Sweden | C2 | 0.654816 | 148.324 | 148.324 | 148.324 | 164.317 | 178.885 | 197.484 | 202.485 |
| 12 | 197.484 | Spain | C2 | 0.512998 | 100 | 100 | 100 | 118.322 | 141.421 | 158.114 | 167.332 |
| 29 | 202.485 | Portugal | C2 | 0.646199 | 114.018 | 114.018 | 114.018 | 130.384 | 158.114 | 178.885 | 192.354 |
| 26 | 187.083 | Netherlands | C2 | 0.660178 | 122.474 | 122.474 | 122.474 | 141.421 | 148.324 | 164.317 | 176.068 |
| 21 | 184.391 | Luxembourg | C2 | 0.618379 | 114.018 | 114.018 | 114.018 | 134,164 | 144.914 | 158.114 | 164.317 |
| 20 | 184.391 | Lithuania | C2 | 0.662973 | 126.491 | 126.491 | 126.491 | 144.914 | 144.914 | 167.332 | 176.068 |
| 22 | 130.384 | Latvia | C2 | 0.574488 | 134.164 | 134.164 | 134.164 | 137.84 | 148.324 | 137.84 | 130.384 |
| 18 | | Iceland | C2 | 0.539233 | 114.018 | 114.018 | 114.018 | 122.474 | 137.84 | 144.914 | 158.114 |
| 13 | | Finland | C2 | 0.662329 | 144.914 | 144.914 | 144.914 | 148.324 | 161.245 | 178.885 | 178.885 |
| 9 | | Denmark | C2 | 0.647816 | 158.114 | 158.114 | 158.114 | 161.245 | 176.068 | 202.485 | 202.485 |
| 3 | | Belgium | C2 | 0.657884 | 130.384 | 130.384 | 130.384 | 134.164 | 151.658 | 161.245 | 170.294 |
| 31 | 122.474 | Serbia | C3 | 0.652859 | 54.7723 | 54.7723 | 54.7723 | 54.7723 | 54.7723 | 70.7107 | 83.666 |
| 19 | 114.018 | Italy | C3 | 0.503911 | 70.7107 | 70.7107 | 70.7107 | 70.7107 | 70.7107 | 83.666 | 94.8683 |
| 11 | 89.4427 | Greece | C3 | 0.666662 | 44.7214 | 44.7214 | 44.7214 | 44.7214 | 44.7214 | 70.7107 | 63.2456 |
| 6 | 130.384 | Cyprus | C3 | 0.646983 | 0 | 0 | 0 | 31.6228 | 54.7723 | 70.7107 | 100 |
| 15 | 109.545 | Croatia | C3 | 0.646792 | 31.6228 | 31.6228 | 31.6228 | 77.4597 | 77.4597 | 83.666 | 89.4427 |



Machine Learning and Predictions

