

THE INFLUENCE OF INCOME INEQUALITIES ON SOCIO-ECONOMIC DEVELOPMENT IN THE EUROPEAN UNION

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Abstract: Recently, there has been observed intensified research on the impact of income inequalities on aspects of socio-economic development in the European Union. However, there are no comprehensive analyses concerning the relationship between these phenomena. Therefore the subject of the paper is the influence of income inequalities on socio-economic development. The author would like to verify the hypothesis that the character of the impact of income inequalities on socio-economic development in the European Union is negative. Analysis was conducted for the European Union in 2004-2017 using the panel data model, also estimated was the synthetic indicator of socio-economic development. The research conducted in the paper leads to ambiguous conclusions. On the one hand, inequalities measured for the whole distribution of income have no influence on socio-economic development in the European Union. However, the income gap between the richest and the poorest hinders the mentioned phenomenon.

Keywords: income inequalities, socio-economic development, economic growth, panel model, Gini coefficient, S80/S20 ratio.

1. Introduction

In recent years one can observe the intensification of interest in the influence of income inequalities on particular aspects of socio-economic development in the European Union, e.g. economic growth, human capital and the natural environment. The impact of income concentration on these categories is generally believed to be negative [Thorbecke, Charumilind 2002; Galor 2000; Persson, Tabellini 1994; Fay 1993; Alesina et al. 2004; Birdsall 1999; Wilkinson 1996; Wildman 2003; Brunner, Marmot 1999; Cole et al. 1992; Hopkins, Kornienko 2010; Van Wilsem 2004; Becker 1993; Shaw, McKay 1942; Boyce 1994]. However, according to the best of this author's knowledge, no attempts to analyze overall relationship between two

mentioned phenomena have been made so far. The only research concerned fragments of this relation, i.e. the influence of income dispersion on the level of education. Moreover, previous studies do not take into consideration the relations between social and economic scopes of analysis. Conducting research from the overall perspective, in regard of all aspects of socio-economic development altogether, could have a crucial impact on the results of analysis.

Therefore, the subject of this paper is the influence of income inequalities on socio-economic development. The author would like to verify the hypothesis that the character of the impact of income inequalities on socio-economic development in the European Union is negative.

Recognition of the character of this influence will enrich economics with knowledge concerning the factors of socio-economic development in the EU. This is crucial in the context of (1) the aim of reaching a high level of this phenomenon and (2) the important initiatives of the World Economic Forum in Davos that wants to alleviate income inequalities, simultaneously promoting socio-economic development [World Economic Forum 2018].

Analysis was carried out for the European Union in 2004-2017. Such length of time series is the consequence of the availability of data in Eurostat from which the information is obtained (this database contains the most complex and high-quality data for the EU member states). Moreover, in 2004 the greatest extension of the European Union to date took place, which resulted in the increase of the quality of statistics in a number of analyzed countries.

Research was carried out using the econometric model. Shortness of time series could be an obstacle in such analysis, so the author decided to employ the panel data model that allows to analyze data both in time and for countries – such a solution increases the number of observations. We considered specifications with fixed effects, random effects and time effects. The paper also estimated the synthetic indicator of socio-economic development. Income inequalities are measured by the Gini coefficient and S80/S20 ratio.

The basis for the econometric analysis is a review of literature concerning the problem of defining and measuring income inequalities and socio-economic development. The author also considered theories on the influence of the former on the latter.

2. Definition and measurement of income inequalities and socio-economic development

2.1. Income inequalities

In this article income inequalities are defined as follows: a situation in which the incomes of society's members (individuals, households or social groups) are not identical [OECD 2011]. Income inequalities are a feature of society that concerns

only the distribution of incomes [Panek 2011]. This paper considered income inequalities between households – this is a popular approach in economic analyses [McKay 2002]. Moreover, the majority of databases offer time series only for this level – statistics for individuals are not available.

Inequalities will be measured on the basis of disposable income. The latter will be understood as “gross income less income tax, regular taxes on wealth, employees', self-employed and unemployed (if applicable) persons' compulsory social insurance contributions, employers' social insurance contributions and inter-household transfers paid”. This definition is proposed by the European Union [2003, p. 208] in association with the Survey of Income and Living Conditions (EU-SILC) that is carried out by all member states of the EU.

Such a definition is the most appropriate as it represents the actual level of income. Additionally, it encompasses all revenues, so differences in the understanding of income sources in various countries are not a problem anymore. In this paper adopting such a definition is also appropriate due to using the Eurostat database – this organization offers the longest reliable time series for European Union member states.

Income that is the basis for the measurement of inequalities should be recalculated in order to ensure comparability in time and between countries. The most popular solutions in this case are Purchasing Power Parity and the equivalence scale. Concerning the former, we used data according to the estimation proposed by Eurostat. As regards the latter, the modified equivalence scale was applied. This is also the solution suggested by Eurostat. This way of transforming income is not appropriate for all member states. However, Eurostat wants to ensure the international comparability of data, therefore it is necessary to use one equivalence scale for all countries [Brandolini 2007].

Different indicators of inequalities are proposed. Some of them are used to measure dispersion in the whole distribution of income (Theil index, Gini coefficient, Atkinson measure). The others could measure the distance between the richest and the poorest (decile ratio, modified quintile ratio, share of income by different income groups). It is necessary to take into consideration both types of indicators [Panek 2011].

The choice of the most appropriate ones should be made on the basis of axioms - features that ought to characterize the measures of income inequalities [Allison 1978; Subramanian 2004; Panek 2011; Amiel, Cowell 1992]. The Gini coefficient seems to be the most useful here. Additionally, it is the most popular (and available in official credible databases) indicator of income inequalities. However, this measure is based on the whole distribution of income, so it is sensitive to outliers [Brzeziński, Kostro 2010], therefore it will be appropriate to consider also an indicator that compares the situation of the richest and the poorest. In this paper, the S80/S20 ratio is adopted. According to Eurostat [2018], “The income quintile share ratio or the S80/S20 ratio is [...] calculated as the ratio of total income

received by the 20 % of the population with the highest income (the top quintile) to that received by the 20 % of the population with the lowest income (the bottom quintile)”. The two mentioned indicators of inequalities give the whole picture of inequalities. Moreover, Eurostat offers the longest time series for these measures.

2.2. Socio-economic development

In the literature there is no compromise on the definition of socio-economic development. The author of the paper tried to conceptualize this phenomenon. On the basis of the literature review (among others: Kupiec [1995], Fritz [2004], Bobrowska and Piasecka [2005], Kompa [2009], Kubiczek [2014]), the author adopted the following definition of socio-economic development: this is a process of qualitative, quantitative, and structural changes. These modifications are consequences of the subjects’ actions that are taken within economic practice. The changes take place in the following fields: *material standard of living* (possibility to satisfy needs associated with consumption of goods and services [economic growth], housing conditions, possibility to make ends meet), *economic structure and entrepreneurship* (investments, employment, innovations, *access to public goods and services* (that results in changes of education level, the way of taking care of someone’s health etc.), *natural environment* (waste, CO₂ emission) and *life satisfaction* (associated with material conditions of life).

The large number of dimensions of socio-economic development is an obstacle in measuring this phenomenon – a common standard in economics is lacking. There are only different indicators proposed that could be treated as attempts to measure socio-economic development. The most popular is the Human Development Index (HDI) that is developed on the basis of three criteria: long and healthy life, knowledge (education), appropriate standard of living [UNDP 2016]. Other indicators based on HDI, stress different aspects of socio-economic development, for example: Gender-Related Development Index, Gender Inequality Index [UNDP 2016], Human Poverty Index [UNDP 1996], Physical Quality of Life Index [UNDP 2012].

These measures allow to describe the level of socio-economic development, nevertheless they have some critical disadvantages. Firstly, they usually encompass only few aspects of the considered phenomenon, however there is a growing belief nowadays that an indicator of socio-economic development should consist of more variables [Milenkovic et al. 2014; Kompa 2009]. Secondly, popular measures concentrate on the material standard of living and financial categories, like income (in HDI it accounts for one-third of the measure). Thirdly, the time series for the proposed indicators are often short and non-continuous which is a major obstacle in econometric analysis.

Because of these disadvantages, the author of the article decided to develop his own synthetic measure of socio-economic development, using the method of

Table 1. Diagnostic variables for a measure of socio-economic development

Indicator	Variable	Coefficient of variation	Type of variable in Hellwig procedure	Influence on socio-economic development
Gross domestic product at market prices; chain linked volumes, percentage change, per capita	v1	208.26%	isolated	S
Final consumption expenditure; chain linked volumes, percentage change, per capita	v2	220.35%	isolated	S
Total population without a bath or a shower, and no indoor flushing toilet in their household	v3	203.43%	satellite	
People living in households with very low work intensity (population aged 0 to 59 years)	v4	34.80%	isolated	D
Average number of rooms per person by type of household and income group from 2003	v5	24.62%	central	S
Inability to face unexpected financial expenses	v6	39.56%	satellite	
Inability to make ends meet (Households making ends meet with great difficulty)	v7	71.80%	satellite	
Total investment to GDP ratio	v8	18.80%	central	S
Employment and activity by gender and age – annual data (as percentage of population)	v9	7.37%		
Unemployment by gender and age – annual data (as percentage of active population)	v10	46.86%	central	D
Long-term unemployment by gender – annual average (as percentage of active population)	v11	73.68%	satellite	
Intramural R&D expenditure (GERD) – Percentage of gross domestic product (GDP)	v12	58.45%	satellite	
Total R&D personnel, as % of total employment	v13	41.71%	satellite	
R&D researchers, as % of total employment		43.79%	satellite	
Final energy consumption (index, 2005=100)	v14	7.06%		
Share of energy from renewable sources in gross final energy consumption	v15	70.37%	isolated	S
Greenhouse gas emissions, base year 1990	v16	31.94%	satellite	
Healthy life years in absolute value at birth – females	v17	7.67%		
Healthy life years in absolute value at birth – males	v18	7.94%		
Self-perceived health (very good)	v19	48.53%	isolated	S
Self-perceived health (very bad)	v20	52.21%	isolated	D
Self-reported unmet needs for medical examination (reason: Too expensive or too far to travel)	v21	91.67%	satellite	
Self-reported unmet needs for medical examination (reason: No unmet needs to declare)	v22	5.72%		
Early leavers from education and training	v23	55.97%	isolated	
Tertiary educational attainment	v24	31.51%	satellite	D
Employment rates of recent graduates	v25	8.24%		
Adult participation in learning	v26	74.90%	satellite	
Young people not in employment or in education and training	v27	33.76%	satellite	
Recorded offences by offence category – police data	v28	81.75%	satellite	
Share of total population reporting problems with crime, violence or vandalism in the area they live	v29	39.7%	isolated	D
Material deprivation rate for the ‘Economic strain’ and ‘Durables’, by number of item of deprivation	v34	94.52%	satellite	
Severe housing deprivation rate	v35	95.98%	central	D

* S – stimulant; D – destimulant.

Source: own estimation.

multidimensional comparison analysis. This indicator encompasses all the aspects that were mentioned in the definition of the phenomenon. Such a measure enables us to analyse socio-economic development better than ever before as it is broader than the indicators developed so far.

A synthetic measure was estimated in four steps – according to the method described by Panek [2007], Guzik, Appenzeller and Jurek [2007], Panek and Zwierzchowski [2013]. First of all, a set of diagnostic variables was chosen, along with merit and formal (statistical) criteria. As concerns the former, the initial set of indicators was designed on the basis of literature review in [Litwiński 2015] and shown in Table 1. This list was verified on the basis of statistical methods: a discrimination analysis and an evaluation of the information capacity of variables. Regarding the former, the value of the coefficient of variation was assessed – variables with a value below 10% were excluded [Panek, Zwierzchowski 2013]. In order to verify the information capacity of variables, we used the Hellwig [1968] procedure that is based on correlation analysis (details are indicated in the paper of this statistician, so there is no need to describe them here). For further estimations, only central and isolated variables were taken.

Secondly, indicators were divided into two groups: stimulants (phenomena that influence socio-economic development positively) and destimulants (phenomena that have a negative influence on socio-economic development). The character of variables was indicated in the last column of Table 1, only indicators that were not rejected in the previous part of analysis were taken into consideration. Destimulants were transformed using a multiplicative inverse that is the most popular method in such cases [Gatnar, Walesiak 2004].

Thirdly, variables were transformed by unitarization:

$$z_{ij} = \frac{x_{ij} - \min_i \{x_{ij}\}}{\max_i \{x_{ij}\} - \min_i \{x_{ij}\}}$$

where: z_{ij} – normalized value of variable j in object i , x_{ij} – original value of variable j in object i , values of indicators that were recalculated in this way range from 0 to 1 and have a more intuitive interpretation.

Fourthly, variables were aggregated using the following formula [Guzik et. al. 2007; Panek, Zwierzchowski 2013]:

$$s_i = \frac{1}{m} \sum_{j=1}^m z_{ij} w_j$$

where: s_i – value of synthetic measure in object i , z_{ij} – normalized value of variable j in object i , w_j – weight for variable j , estimated according to procedure BVP, described by Panek [2007].

Such an aggregation formula is the most popular in economic research that engages methods of comparison analysis [Panek, Zwierzchowski 2013].

The indicator calculated according to the procedure that was described above covers the range $<0;1>$. The higher the value of a measure, the higher the level of socio-economic development.

3. Previous studies of the influence of income inequalities on aspects of socio-economic development

The research presented in the literature so far allowed to confirm that income inequalities influence the following aspects of socio-economic development (understood in the way that was described in the previous part of the article): (1) economic growth [Thorbecke, Charumilind 2002; Galor 2000; Persson, Tabellini 1994; Fay 1993], (2) education [Alesina et al. 2004; Birdsall 1999], (3) health (physical and mental welfare) [Wilkinson 1996; Wildman 2003; Brunner, Marmot 1999], (4) life satisfaction [Cole et al. 1992; Hopkins, Kornienko 2010], (5) crime [Van Wilsem 2004; Becker 1993; Shaw, McKay 1942], (6) natural environment [Boyce 1994].

Table 2. Quantitative analyses of the influence of income inequalities on economic growth

Author of the research	Method of analysis	Number of countries	Conclusions regarding the influence of income inequalities on socio-economic development
[Alesina, Rodrik 1994]	Cross-sectional regressions	70	Income inequalities have a negative influence on socio-economic development
[Perotti 1994]		71	
[Persson, Tabellini 1994]	Panel model	9	
[Knack, Keefer 1997]	Cross-sectional regressions	56	Income inequalities do not have an influence on socio-economic development
[Perotti 1994]		67	
[Benhabib-Spiegel 1994]		40	
[Deininger, Squire 1998]		46	
[Li, Zou 1998]	Panel model with fixed effects	46	Income inequalities have a positive influence on socio-economic development
[Forbes 2000]		45	
[Milanovic 2002]		24	Income inequalities have a negative influence on socio-economic development
[Tanninen 1999]	Cross-sectional regressions	52	Income inequalities have a negative influence on socio-economic development
[Knell 1999]		83	
[Rehme 2002]		34	Income inequalities have a negative influence on socio-economic development in high-income countries
[Ostry, Berg, Tsangarides 2014]		153	Alleviation of income inequalities has a positive influence on economic growth
[Barro 2000]	Panel model with random effects	46	Income inequalities have a negative influence on socio-economic development in low-income countries and a positive effect in high-income countries

Source: literature review on the basis of the indicated references.

The analyses of the relationship between income inequalities and socio-economic development that are listed above do not have an overall character; there was research on particular fragments of the scientific scope of relation that is the subject of this paper, e.g. explaining only the influence of income inequalities on economic growth or the meaning of income dispersion for the distribution of educational opportunities. Therefore the studies mentioned above have rarely taken into account the interactions between social and economic aspects (elements) of different aspects of the socio-economic development. Moreover, studies on the influence of income inequalities on the social dimensions of the latter (e.g. health, life satisfaction) are conducted using mainly qualitative methods.

In the context of this paper it is useful to review the quantitative analyses that have been carried out so far in order to obtain knowledge about the methods which were used by other scientists. The research based on econometric tools is presented mainly in papers concerning the influence of income inequalities on economic growth. The results of these studies are shown in Table 2. On the basis of the majority of analyses, it could be realized that income inequalities have a negative influence on economic growth. The most popular method of research was regression for cross-sectional data. The author of this paper decided to use a more advanced tool, the panel model, analyzing the relationship between income inequalities and a category broader than economic growth – socio-economic development.

4. Empirical analysis of the influence of income inequalities on socio-economic development

The empirical research of relationship that is the subject of this paper was conducted for the European Union in the years 2004-2017. The variables that were taken into account are the synthetic indicator of socio-economic development which was estimated in the previous part of the article (dependent variable) and measures of income inequalities: the Gini coefficient and S80/S20 ratio (independent variables).

For each indicator of income dispersion a separate model was estimated. Such a solution is a consequence of the fact that both measures represent the same phenomenon, so they contain similar information i.e. the correlation coefficient between them (in the EU in the period 2004-2017) is 0.86. Additionally, the estimation of two separate models allows to indicate the aspect of income inequalities (dispersion for the whole distribution or the gap between the richest and the poorest) that has a stronger effect on socio-economic development.

In the paper we used the following abbreviations of variables' names: **development** – socio-economic development, **Gini** – Gini coefficient, **S80/S20** – modified income quintile share ratio. The models are estimated using Gretl software. In all the conducted tests the significance level is equal to 0.05.

The method that was used to recognize the character of influence of income inequalities on socio-economic development was panel data analysis. However before estimating the model, the data should be verified in order to ensure that it is of good quality. One of the serious problems in the time series could be unit root (non-stationarity of time series) that is often a reason for spurious regression, a situation in which the model seems to be correct but the relation between the analyzed phenomena does not exist in economic practice.

A non-stationary time series could be generated by a model of random walk with drift or a model with trend. The kind of model for a particular variable is the basis for the decision on the way of transforming the variables. In the case of random walk, differences should be calculated first, while the second model requires removing the trend [Charemza, Deadman 1997; Kusideł 2000; Maddala 2008; Nelson, Kang 1984]. The author decided to use the filter proposed by Hodrick and Prescott [1997].

The source of unit root was recognized on the basis of the test proposed by Dickey and Fuller [1981]. Within the procedure, we estimated the following model:

$$\Delta y_t = \alpha_0 + \delta \alpha_1 t + \delta y_{t-1} + \sum_{i=1}^k \gamma_i \Delta y_{t-i} + \epsilon_t,$$

where: Δy_t – first difference; equal to $y_t - y_{t-1}$, y_{t-i} – value of variable in the period $t-i$, t – time variable, α_0 , α_1 , δ , γ_i – parameters.

In such an equation, stochastic and deterministic trends are taken into account. Therefore the null hypothesis states that time series is generated by a model of random walk with drift ($\delta = 0$), while the alternative hypothesis says that time series is a realization of a model with trend ($\delta < 0$). Test statistics have a distribution proposed by Dickey and Fuller [1981] and are calculated according to the formula suggested by the authors.

The results of the above procedure are presented in Table 3. All the variables are generated by the model with trend. The time series were transformed using the Hodrick-Prescott filter.

Table 3. Results of the ADF test for cross-sectional data from the EU in 2004-2017

Variable	Number of lags (k)	p-value	Conclusions on null hypothesis	Conclusions on the model that generates time series
development	2	0.00	Reject	Model with trend
Gini	1	0.00	Reject	Model with trend
S80/S20	3	0.00	Reject	Model with trend

Source: own estimation.

The choice of econometric method that was used to recognize the character of the influence of income inequalities on socio-economic development in the European

Union, the panel model, was the result of review of analyses conducted so far (Table 2). The analysis of cross-sectional data is some kind of standard in research on the effect of income dispersion on one of the aspects of socio-economic development, economic growth. Besides, the panel model is a useful econometric tool in the analysis of short time series (as in this paper), because it allows to take into account cross-sectional data, data for different objects (mainly countries or regions) and in time. Increasing the number of observations enables the researcher to take into consideration the relations between the objects. Additionally, thanks to rise of dataset, dispersion is growing, which decreases the risk of collinearity, moreover estimates of the parameters are more accurate [Muszyńska 2006; Maddala 2008].

In this paper we considered four types of panel model: with fixed effects, with fixed and time effects, with random effects, with random and time effects. Each model, regardless of the specification, is expressed by the formula:

$$y_{it} = \alpha + \beta x_{it} + u_{it}$$

while:

$$u_{it} = \mu_i + \lambda_t + v_{it}$$

where: y_{it} – dependent variable in object i and period t , x_{it} – independent variable in object i and period t , α, β – parameters of a model, u_{it} – random component, μ_i – individual effect for object (country), λ_t – time effect, v_{it} – the rest of random component, generated by a purely random process.

The model with fixed effects is based on the assumption that individual effects μ_i are constant parameters for the particular analyzed object [Baltagi 2001]. In order to check if such specification is correct, the Breusch-Pagan test is used. The null hypothesis of this procedure says that a variation of individual effects is equal to 0 ($\mu_1 = \mu_2 = \dots = \mu_T = \mu$, czyli $\sigma_\mu^2 = 0$). Rejection of this hypothesis means that the model is appropriate for the analyzed dataset [Maddala 2008].

In the model with random effects, it is assumed that individual effect μ_i is generated by a random variable with known distribution [Baltagi 2001]. The test proposed by Hausman allows to check if such specification is correct. It has null hypothesis about the lack of correlation between individual effects μ_i and independent variable x_{it} . Rejection of this hypothesis means that the model should not be used.

An additional specification for both types of model is the introduction of time effects. In this class of models, time effects λ_t are constant for each object in a particular period. In order to check if time effects are necessary, we could use the Wald test with null hypothesis saying that the dummy variables for time are insignificant. Rejection of this hypothesis means that the introduction of time effects is justified [Maddala 2008].

The quality of the model that was chosen on the basis of the mentioned procedures was verified by two tests. Firstly, the significance of the independent

variable was tested, using the procedure proposed by Gosset [1908] (Student t-test) with null hypothesis about the lack of influence of the independent variable on the dependent variable. Secondly, the test of Ljung and Box [1978] was employed. This procedure has null hypothesis saying that the random component is not autocorrelated.

The results of the tests that allow to choose the correct specification of the model for the synthetic indicator of socio-economic development and the Gini coefficient are presented in Tables 4 and 5. The model with random effects and time effects is the most appropriate here. This is confirmed not only by the results of the respective tests but also by the values of information criteria – there were analyzed indicators proposed by Akaike [1974], Hanan-Quinn [1979] and Schwarz [1978].

Table 4. Results of the Breusch-Pagan and Hausman tests for a synthetic indicator of socio-economic development and the Gini coefficient

		Model with fixed effects	Model with random effects
Test		Breusch-Pagan	Hausman
p-value		0.09	0.36
Conclusion about H0		Accept	Accept
Inf. criteria	Akaike	-2396.60	-2433.39
	Hanan-Quinn	-2350.96	-2430.25
	Schwarz	-2281.44	-2425.45
Conclusion about a model		INAPPROPRIATE	APPROPRIATE

Source: own estimation.

Table 5. Results of the Wald tests for a synthetic indicator of socio-economic development and the Gini coefficient

		Model with time effects	
Test		Wald	
p-value		1.89E-87	
Conclusion about H0		Reject	
Inf. criteria		For model with random effects without time effects	For model with random effects and time effects
	Akaike	-2433.39	-2708.91
	Hanan-Quinn	-2430.25	-2685.30
	Schwarz	-2425.45	-2649.34
Conclusion about a model		INAPPROPRIATE	APPROPRIATE

Source: own estimation.

Table 5 contains the outcomes of the estimation of the panel model for the Gini coefficient and an indicator of socio-economic development. According to the results, the parameter for the independent variable is insignificant. Income inequalities, measured for the whole distribution of income, do not influence socio-economic development in the European Union in 2004-2017.

Table 6. Results of estimation of the panel model for a synthetic indicator of socio-economic development and the Gini coefficient in the European Union in 2004-2017

Criterion		Estimations	
Parameters		Constant	Gini
	Estimation	0.04	-3,42E-05
Student t-test	p-value	0.75	
	Conclusion about H0	Accept	
	Conclusion about influence of independent variable	Income inequalities, measured by the Gini coefficient, do not influence socio-economic development. measured by synthetic indicator	
Ljung-Box autocorrelation test	p-value	0.36	
	Conclusion about H0	Accept	
	Conclusion about autocorrelation	Random component is not autocorrelated	

Source: own estimation.

Tables 7 and 8 contain the results of the tests that allow to choose the correct model for a synthetic indicator of socio-economic development and S80/S20 ratio. In this case the specification with random effects and time effects is also appropriate for the analyzed data. The Breusch-Pagan, Hausman and Wald tests

Table 7. Results of the Breusch-Pagan and Hausman tests for a synthetic indicator of socio-economic development and S80/S20 ratio

		Model with fixed effects	Model with random effects
Test		Breusch-Pagan	Hausman
p-value		0.09	0.39
Conclusion about H0		Accept	Accept
Inf. criteria	Akaike	-2396.53	-2433.38
	Hanan-Quinn	-2350.88	-2430.24
	Schwarz	-2281.36	-2425.44
Conclusion about a model		INAPPROPRIATE	APPROPRIATE

Source: own estimation.

Table 8. Results of the Wald tests for a synthetic indicator of socio-economic development and S80/S20 ratio

		Model with time effects	
Test		Wald	
p-value		1.33E-87	
Conclusion about H0		Reject	
Inf. criteria		For model with random effects without time effects	For model with random effects and time effects
	Akaike	-2433.38	-2708.91
	Hanan-Quinn	-2430.24	-2685.30
	Schwarz	-2425.44	-2649.34
Conclusion about a model		INAPPROPRIATE	APPROPRIATE

Source: own estimation.

enable us to say that the model is correct, also the values of the Akaike, Hanan-Quinn and Schwarz criteria indicate that the loss of information is the lowest for this specification.

Table 9. Results of estimation of the panel model for a synthetic indicator of socio-economic development and S80/S20 ratio in the European Union in 2004-2017

Criterion		Estimations	
Parameters		constant	Gini
	Estimation	0.04	-0,0002
Student t-test	p-value	0.01	
	Conclusion about H0	Reject	
	Conclusion about influence of independent variable	Income inequalities, measured by S80/S20 ratio, influence socio-economic development. measured by synthetic indicator	
Ljung-Box autocorrelation test	p-value	0.54	
	Conclusion about H0	Accept	
	Conclusion about autocorrelation	Random component is not autocorrelated	

Source: own estimation.

The results of the estimation of the panel model for measuring socio-economic development and S80/S20 ratio are presented in Table 9. On the basis of Student's t-test, it could be said that income quintile share ratio influences significantly the indicator of socio-economic development. Thus the income gap between the richest and the poorest (contrary to inequalities measured for the whole distribution) affects socio-economic development in the European Union in the years 2004-2017. The character of this influence is negative as the concentration of incomes hinders socio-economic development.

The model is of good quality. Firstly, the values of information criteria are low. Secondly, according to the Ljung-Box test, random component is not autocorrelated.

5. Conclusions

The analysis presented in the paper leads to ambiguous conclusions. On the one hand, income inequalities measured for the whole distribution of income have no influence on socio-economic development in the European Union. However, income gap between the richest and the poorest hinders the mentioned phenomenon.

Thus the hypothesis of the article should not be accepted unequivocally. Only the differences between groups with high and low incomes could be said to have a negative impact on socio-economic development in the EU.

Such conclusions about the character of influence have their limitations. The results of econometric analysis depend on the chosen econometric tool and the

variables that are taken into account. Therefore the study presented in the paper could be treated only as an attempt to recognize the character of impact of income inequalities on socio-economic development in the European Union.

Further research could identify elements of socio-economic development that are affected by income inequalities more strongly than the others, also recognizing the period in which the time effect in the panel model appeared.

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WPLYW NIERÓWNOŚCI DOCHODOWYCH NA ROZWÓJ SPOŁECZNO-EKONOMICZNY W UNII EUROPEJSKIEJ

Streszczenie: W ostatnich latach prowadzi się liczne analizy ilościowe wpływu rozwarstwienia dochodowego na aspekty rozwoju społeczno-ekonomicznego w krajach rozwiniętych. Brakuje jednak badań, które miałyby charakter kompleksowy. Wnioski z analizy tego typu mogłyby być istotną podstawą kształtowania polityki gospodarczej. Dlatego też problemem podjętym w opracowaniu jest wpływ nierówności dochodowych na poziom rozwoju społeczno-ekonomicznego. Cel artykułu stanowi weryfikacja hipotezy o negatywnym charakterze wspomnianej zależności w Unii Europejskiej. Analiza została przeprowadzona dla krajów Unii Europejskiej w latach 2004-2017 za pomocą modelu panelowego. Badanie prowadzi do niejednoznacznych rezultatów. Z jednej strony nierówności mierzone dla całego rozkładu dochodów nie mają wpływu na rozwój społeczno-ekonomiczny w Unii Europejskiej, z drugiej strony dystans między najbogatszymi i najbardziej ubogimi wpływa hamująco na ten proces.

Słowa kluczowe: nierówności dochodowe, rozwój społeczno-ekonomiczny, wzrost ekonomiczny, model panelowy, współczynnik Giniego, wskaźnik S80/S20.