Relations between the inflow of foreign direct investments and economic indicators in OECD countries

**Ladislav SUHÁNYI**, **Alžbeta SUHÁNYIOVÁ**, **Jaroslav KOREČKO**, **Lucia BEDNÁROVÁ**, **Jaroslava KÁDÁROVÁ**, **Hamdi DERKAWI** and **Kamila BAČOVÁ**

**Abstract**

The importance of foreign direct investment in the mining industry can be illustrated in the case of countries whose economy relies heavily on the mining sector as this attracts major foreign investment to them. The aim of the study is to determine whether there is a link between the inflow of foreign direct investment and the selected macroeconomic indicators in OECD countries and to identify groups of countries showing similar characteristics in this context. The evaluation of the interconnection of selected indicators in this study was made firstly by examining if there is a link between the examined macroeconomic indicators (independent variables) and the FDI inflow (dependent variable). The results showed that an increase in labour productivity could be reflected in an increase in FDI inflows. Based on the results obtained, the backward link in relation to labour productivity was also examined, whether there is a link between the macroeconomic indicators (independent variables) and the labour productivity (dependent variable), resulting in that an increase in FDI inflows could translate into an increase in labour productivity. The paper also included cluster analysis, the main goal of which was to determine countries based on the evaluation of labour productivity and the inflow of foreign direct investment. Four clusters have been identified, two large ones and two clusters containing individual countries (Ireland and Luxembourg). All of the V4 countries are included in one of the large clusters, including the countries which are characterised by low variability in average labour productivity and high variability in the average inflow of foreign investment. The mining industry’s recommendations, conclusions, and implications are focused mainly on this cluster of countries.

**Keywords**

Foreign direct investment, labour productivity, economic indicators, OECD countries, clusters.
Introduction

Foreign direct investments (FDI) are an integral part of an open and efficient international economic system and a major catalyst for development. At the same time, the benefits of FDIs are not growing automatically and even equally in all countries, sectors or local communities. Attracting FDIs to countries and reaping the full benefits of FDI depends on national policies and the architecture of international investment. This is a challenge, especially for host countries, which need to put in place a clear, broad-based, and efficient environment for investment and for building human and institutional capacity to put it into practice (OECD, 2002).

Theoretical discussions on the impact of foreign direct investments on economic growth date back to the MacDougall (1960) period. In recent years, the impact of FDI in various countries and economies has received a great deal of attention in both theoretical and empirical literature (Inekwe, 2013; Imoudu, 2012; Adams, 2009; Akinlo, 2004).

At the aggregate level, FDI is generally accepted as a key factor supporting economic growth. For example, Han and Wei (2015), who used a regression tree approach with 17 potential regressors for a panel of individual countries, found that foreign direct investment is one of the most important factors in explaining growth performance, especially for middle-income countries.

The inflow of foreign direct investment into the country is supposed to positively affect the development of the gross domestic product, labour productivity, and unemployment rate by creating space for new jobs, new technologies and new production processes, and know-how, contributing to increasing domestic qualification human capital. In the scientific literature can be found a number of analyses of the impact of foreign direct investment on economic indicators, such as gross domestic product, labour productivity, unemployment rate.

Theoretical review

Foreign direct investment (FDI) is seen as a means of transmitting new ideas, advanced techniques, technologies and skills across borders, creating significant spillover effects (Kumar & Russel 2002, Campisi et al. 2013).

Motives for foreign direct investments:

- According to the traditional view, the natural (perverse) direction of capital flow is from the rich country to the developing country. The motivation for this is the higher marginal productivity of capital.
- According to eclectic theory (Dunning 1980), the necessary conditions for the realisation of FDI (OLI paradigm) are the following:
  - Ownership specific advantages: competitive advantages of the company (e.g. technology, management & marketing skills, R&D, company size, etc.),
  - Location-specific advantages: e.g. natural resources, cheap labour, state aid, etc.,
  - Internalisation advantages: if the benefits of FDI management are higher than licensing, then it is more attractive to perform the value chain activity in-house than to have it performed by an external party,

while in eclectic theory, the underlying motives of FDI include gaining resources, gaining market share, increasing efficiency, and gaining strategic advantage.

The results of various researches and studies point to the positive effect of foreign direct investment on economic growth expressed by the growth of the gross domestic product, on the growth of labour productivity, on the growth of employment, the decline in unemployment, and also on the growth of wages.

This paper focuses on the possible link between FDI and between GDP, wages, employment and labour productivity.

However, research results do not always clearly confirm the mentioned positive effect of FDI. The research of Bačová (2016) showed that in Slovakia in 2009–2015, there was only a small positive association between FDI inflows and between GDP growth, labour productivity growth, employment; a moderate positive association between FDI inflows and the unemployment rate; and a moderate negative association between FDI inflows and the average nominal wage. The effect of FDI on GDP was also examined by Björn and Han (2017), who found that, in general, a significant positive effect of FDI on GDP could not be demonstrated, but in the case of developing Asian countries, FDI significantly improves GDP growth. Wang and Wei (2017) examined the relationship between exchange rates, economic growth and foreign direct investment. According to their conclusions, economic growth is slowing down due to foreign direct investment, but economic growth will return to its original level in the next period.

In theory, there are several potential ways in which FDI can cause growth (Kosztowniak 2016). Macroeconomic analysis generally supports a positive connection between FDI and growth (Borensztein et al. 1998, Kottaridi et al. 2010), a connection which is further reinforced in countries with a well-developed financial market (Alfaro et al. 2004). The Solow-type standard neoclassical growth models (Solow 1999) suggest that FDI
increases the capital stock and thus promotes growth in the host economy by financing capital formation (Brems, 1970; Kida, 2014). FDI promotes economic growth with a short run effect by increasing the volume of investment and/or its efficiency (Kosztowniak, 2016).

Overall conclusions presented by the previously mentioned research of the authors Biørn and Han (2017) are: (1) Aggregate FDI has a positive, but insignificant effect on aggregate GDP based on the full country panel; (2) for the developing Asian countries, FDI significantly improves GDP growth; and (3) manufacturing FDI impacts both manufacturing and service GDP growth positively.

The author Kosztowniak (2016) confirmed the bi-directional relationships between FDI and GDP in Poland. However, the impact of GDP on attracting FDI inflows to Poland is stronger than that of FDI on GDP growth. Thus, the Polish developmental policy should concentrate on three essential determinants (pillars) of growth: employment growth, attracting FDI (with emphasis on improvement in the type of inflowing investment), and increasing domestic the value and productivity of domestic investment. According to Kasych (2019), cross-border cooperation within one company can help with company management. In a way, it is also an investment, but it does not have to be just about financial assistance. Kasych and Medvedeva (2019) also add that cross-border transfer of new technologies can also help.

The level of wages is also an indicator that can be in relation to the FDI flow. The study of Halmos (2011) showed positive and significant relations between increasing income inequality and increasing level of FDI in Eastern-European states, although the relation between FDI inflow and GINI index was not demonstrable. The study of Bacovic et al. (2021) revealed that the growth in wages has a negative impact on FDI inflows in the EU countries, but their research also discovered a paradoxical scenario in a specific case of Balkan countries, where increased wages actually have a positive impact on FDI inflows.

The impact of foreign direct investment on overall wages depends not only on how high the foreign firm is. A cross-sectional study by Lipsy and Sjöholm (2004) made on Indonesian manufacturing industries and provinces implies that a higher foreign presence raises the general wage level. The FDI can influence the wage level in different industries. The results of Li et al. (2018) showed that FDI in the generalised virtual economy industry could promote the increase of average wage level in the way that every 1% increase of FDI increases the average wage level in the industry by 0.88%. From the other perspective, the authors Damijan and Kostevc (2011) examined the role of FDI in the adjustment pattern of regional wages, and the results showed that in most cases, the FDI had contributed significantly to faster adjustment of relative regional wages.

Experience shows that governments worldwide seek to attract foreign direct investments and support them with trade policy incentives, financial assistance and tax incentives (Blomstrom and Kokko, 2003). One of the important motivations of governments is the assumption that foreign companies stimulate productivity and employment either directly through their own higher productivity and employment growth or through an indirect side effect (Blomstrom and Kokko, 1998; Görg and Strobl, 2001; Girma, 2005). Inekwe (2013) has conducted a study examining the links between Nigerian employment and foreign direct investment (FDI) in the manufacturing and servicing sectors between 1990 and 2009. Based on the results can be concluded that in that case, FDI in the manufacturing sector has a positive relationship with employment rate while FDI in the servicing sector has a negative relationship with employment rate (in both cases, unidirectional causality from FDI to employment rate was shown). The results are supported by another research of Waldkirch et al. (2009) examining the relation between FDI and employment data in Mexico, which showed that FDI has a significantly positive, though the quantitatively modest impact on manufacturing employment, and they also concluded that the employment enhancing effects of FDI are larger in export-oriented industries. Based on Caves (2007), foreign acquisitions lead to a reduction in labour-use inefficiency, which is consistent with the internalisation theory of FDI, which postulates that multinational firms transfer a range of intangible proprietary assets to their affiliates. This statement is consistent with the empirical estimates of Girma (2005), who, on the other hand, in addition to the long-time effect, also found out that the gain in efficiency is accompanied by negative as well as positive employment effects, suggesting that foreign investments can create as well as destroy jobs in short to medium run. Haskova and Volf (2017) state that foreign direct investment has a positive effect on the employment and GDP of the state, especially in the region in which the investment was made. Therefore, differences are in existence regarding the GDP levels of the regions of the Czech Republic (Civelek et al., 2019). Local currencies might be an effective solution to minimise the inequalities in economic conditions of various geographical regions (Kļjučnikov et al. 2020a; Kļjučnikov et al., 2020b; Kļjučnikov et al., 2020c).

The indications of the literature review led us to examine also labour productivity (for the purposes of this study meaning the share of GDP per employed persons) which was also used by Bacovic et al. (2021) in their study on the determinants of FDI inflows. Technological externalities/spillovers from FDI may increase productivity in general (Lipsy and Sjöholm, 2004), but there are also other studies showing that inflows of FDI might also lead to a lower scale of production and lower productivity in domestic firms (Aitken and Harrison 1999). The study of Bacovic et al. (2021) examined the labour productivity as a determinant of FDI inflows focusing on Balkan countries in comparison with EU countries, and they found out that, in Balkan countries, FDI inflows respond negatively to shocks in labour productivity, which is contrary to the experience of EU countries.
According to Lavíčková, Kolafová and Turínská (2021), it is possible to increase employee productivity through free language courses paid for by their employer.

**Material and Methods**

The aim of the study is to determine whether there is a link between the inflow of foreign direct investment and the selected macroeconomic indicators in OECD countries and to identify groups of countries showing similar characteristics in this context.

The research sample consisted of 36 OECD countries. These are countries that are considered to be more developed countries. The data available from the OECD database for the years 2005-2019 were adjusted and processed to obtain the final indicators described below. US dollars in current PPPs were used for source data in monetary units.

The indicator of employment (EM) was calculated as the number of employed people (not only employees) per 100,000 inhabitants. The indicator of labour productivity (LP) was calculated as GDP per employed person. The indicator of average wage (AW) was calculated as a yearly average wage of an individual (single person at 100% of average gross earnings before taxes with no child). The indicator of GDP was used in absolute numbers per head.

FDI-related indicators can be (Dombi, 2015):
- Flow indicators – the value of a foreign direct investment in a country (FDI imports) or from a country (FDI exports) during a given period
- Stock indicators – the cumulative value of FDI imports and exports up to a given date (imported and exported FDI stocks)

In this paper, for the indicator of foreign direct investments (FDI) the total FDI inward financial flows per head were calculated.

The evaluation of the interconnection of selected indicators in this study was based on the process of validation of the following working hypotheses:

Hypothesis 1: There is a link between the examined macroeconomic indicators (independent variables) and the FDI inflow (dependent variable).

After examining Hypothesis 1 and based on the results obtained, a hypothesis was formulated in order to verify the backward link in relation to labour productivity:

Hypothesis 2: There is a link between the examined macroeconomic indicators (independent variables) and labour productivity (dependent variable).

The proposed “Primary Model of FDI” reflects the relationship between all the mentioned independent variables and the dependent FDI variable. The explanatory macroeconomic variables were employment (EM), average wage (AW), labour productivity (LP) and gross domestic product (GDP) in the OECD countries. The aim of using the regression model was to determine the expected effects of selected explanatory variables on the explained variable. The intention was to find and create the best linear model expressing the dependencies between the selected variables. The model and the expected effects of the explanatory variables on the explained variable were formulated as follows (Meixnerová & Krajháč, 2020):

\[ FDI_t = f(EM_t, AW_t, LP_t, GDP_t) \] (1)

\[ FDI_t = f(+, −, +, +) \]

The proposed “Primary Model of LP” reflects the between all the following independent variables on the dependent LP variable:

\[ LP_t = f(EM_t, AW_t, FDI_t, GDP_t) \] (2)

\[ LP_t = f(−, +, +, +) \]

All “primary” and “secondary” regression models were tested for the presence of adverse events in the model. The “secondary” regression models were derived from the “primary” regression models replacing the values of the indicators by their differences. The presence of heteroskedasticity was tested by the Breusch-Pagan test, the autocorrelation was verified by the Durbin-Watson test, the presence of multicollinearity was verified by the VIF (variance inflation factor) test.
The last part of the analysis was devoted to the application of cluster analysis. The main goal was to determine countries based on the evaluation of labour productivity and the inflow of foreign direct investment. Cluster analysis (CLU) is useful in cases where objects have a natural tendency to cluster. The objectives of cluster analysis cannot be separated from searching and selecting the suitable characters to define the clustered objects. The found clusters describe the data structure only with respect to the selected characters (Meloun, 2012). The choice of features of objects must be preceded by an analysis of theoretical and practical criteria for their justification. Therefore, the cluster analysis was preceded by relevant statistical research, even in our case.

The goal of the CLU is to cover a set of objects with their subsets, which may not be disjoint. After identifying significantly different groups of objects, it is possible to concretise them. The groups may differ, for example, by the level of the monitored trait (variable) or its variability (Kráľ, 2009).

Since practical data mining problems high-dimensional data are clustered, the resulting clusters are high-dimensional geometrical objects which are difficult to analyse and interpret. A low-dimensional graphical representation of the clusters could be much more informative than such a single value of the cluster validity. One can cluster by eye and qualitatively validate conclusions drawn from clustering algorithms (Abonyi & Feil, 2007).

In this paper, both hierarchical and non-hierarchical clustering were used.

- **Hierarchical clustering**

  The most commonly used measure of the distance of objects is the Euclidean distance, respectively geometric distance. The Euclidean distance forms the basis of Ward’s clustering method, which have been used in the presented model. This is the so-called divisional clustering, which is based on the set of all objects (countries) as a single cluster and its gradual division leads to a system of clusters. The advantage of hierarchical methods is that it is not necessary to know in advance the number of clusters before the clustering process itself.

  Euclidean distance is defined by the formula:

  \[
  d_{ij} = \sqrt{\sum_{k=1}^{K} (x_{ik} - x_{jk})^2}
  \]  

  Where \(x_{ik}\) is the value of \(k\) variable for \(i\)-th object and \(x_{jk}\) is the value of \(k\) variable for \(j\)-th object. The rule of linking statistical units into clusters is then determined for calculated distance.

  Ward’s method is a “procedure for forming hierarchical groups of mutually exclusive subsets on the basis of their similarity with respect to specified characteristics” (Ward, 1963).

  The principle of Ward’s clustering method is to minimise the heterogeneity of clusters according to the criterion of the minimum increment of the intra-group sum of squares of deviations of objects from the centre of gravity of clusters. If the cluster consists of \(j\) objects that are characterised by \(m\) characters, a matrix \(j \times m\) with elements \(x_{ik}\) (value of the \(k\)-th character for the \(j\)-th object) is available. Within-cluster variability (sum of squares within – \(SS_W\)) is then given by:

  \[
  SS_W = \sum_{k=1}^{m} \sum_{i=1}^{j} (x_{ik} - \bar{x}_k)^2 \]  

  while

  \[
  \bar{x}_k = \frac{1}{j} \sum_{i=1}^{j} x_{ik}
  \]

  Adding more clusters with \(j_i\) objects increases the number of rows of the original matrix to \(j + j_i\), and \(SS_W\) counts for a larger number of objects.

- **Non-hierarchical clustering**

  Non-hierarchical methods use the optimisation procedure, where it happens that during the formation of clusters, an object (country) is closer or further away from the cluster in which it is currently located. Then the optimisation procedure places it in another (closer) cluster. A key problem with non-hierarchical methods is the choice of the number of clusters in advance. For this reason, hierarchical clustering was first carried out in this research, which indicated the number of clusters in the set of surveyed countries. The same number was used in this case as well. Later, this number was optimised according to the position of the objects.

  In the subsequent clustering process, the K-means method was used, which consists in dividing \(n\) objects with \(m\) characters into \(k\) clusters so that the between-cluster sum of squares (sum of squares between – \(SS_B\)) is minimised.

  \[
  SS_B = \frac{nm}{nm-m} \sum_{i=1}^{n} \sum_{j=1}^{m} \sum_{k=1}^{n} (1 - \delta_{ij})(y_{ij} - c_{ik})^2
  \]
It is assumed that $n$ objects are divided into $k$ clusters. Then the $k$-th cluster contains $n_k$ objects. Each object is described by $m$ characters. The missing value of the $i$-th character in the $j$-th line and in the $k$-th cluster is denoted as $\delta_{ijk}$. The $x_{ij}$ data are pre-standardised and denoted as $y_{ij}$. The value of $c_{ik}$ is the mean value (average) of the $i$-th character in the $k$-th cluster (Meloun, 2012).

**Results**

As mentioned above, we will present the implementation and results of stepwise regression in the following section. The estimated vectors of regression coefficients in the model generate the influence of the explanatory variables on the explained variable at the significance level $\alpha$. The performed test of statistical significance of variables determined which explanatory variables we will consider as significant determinants of FDI and LP. The above facts illustrate the following outputs:

**Tab. 1. Estimation of regression coefficients in Primary Model of FDI**

|                | Estimate | Std. Error | t value | Pr(>|t|) |
|----------------|----------|------------|---------|----------|
| (Intercept)    | -3.917e+04 | 1.433e+04  | -2.734  | 0.00646  ** |
| EM             | 7.830e-01  | 3.223e-01  | 2.429   | 0.01545  * |
| LP             | 5.451e-01  | 1.754e-01  | 3.107   | 0.00199  ** |
| AW             | -1.731e-01 | 5.631e-02  | -3.075  | 0.00221  ** |
| GDP            | -8.758e-01 | 3.974e-01  | -2.204  | 0.02797  * |

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 1

Residual standard error: 10630 on 535 degrees of freedom
Multiple R-squared: 0.1099, Adjusted R-squared: 0.1033
F-statistic: 16.52 on 4 and 535 DF, p-value: 9.029e-13

Source: authors

Testing the “Primary Model of FDI” showed that this regression model was burdened by heteroskedasticity and autocorrelation of variables (Table 1). Due to the requirement to remove undesirable phenomena from the original model, the variables in the “Secondary Model of FDI” have therefore been replaced by their differences.

**Tab. 2. Final estimation of regression coefficients in Secondary Model of FDI**

|                | Estimate | Std. Error | t value | Pr(>|t|) |
|----------------|----------|------------|---------|----------|
| (Intercept)    | -5.10917 | 607.38898  | -0.008  | 0.9933   |
| diff(LP)       | 0.07895  | 0.03706    | 2.131   | 0.0336   * |

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 1

Residual standard error: 14100 on 537 degrees of freedom
Multiple R-squared: 0.1099, Adjusted R-squared: 0.1033
F-statistic: 16.52 on 4 and 535 DF, p-value: 0.03358

Source: authors

Based on the testing, exogenous variables were removed from the “Secondary Model of FDI”, which showed statistical insignificance according to the achieved $t$-values. Low $t$-values indicated a weak explanatory power of the coefficient of the given variable. According to the model, the variable AW is statistically insignificant. In other modifications of the model, this variable was disregarded. In the “Secondary Model of FDI”, the variables had a normal distribution, but the multicollinearity test showed its presence in the model. Naturally, the pairs of variables GDP and LP, as well as the EM and LP, show a high degree of mutual correlation. Based on their mutual correlation, it was possible to omit these variables. According to the estimators and $t$-values in the model, it was optimal to omit GDP and EM first. The final estimation of regression coefficients at the significance level $\alpha = 0.05$ shows that if the labour productivity in OECD countries increases by one unit, the foreign investment flows will increase by 0.07895 units (Table 2). This result cannot be quantified in this way entirely, but it appears that there is a link between FDI flows and labour productivity that needs to be taken into account in decision-making. Thus, based on the findings, an increase in labour productivity could be reflected in an increase in FDI inflows.
Tab. 3. Estimation of regression coefficients in Primary Model of LP

| Residuals:                  | Min  | 1Q   | Median | 3Q   | Max  |
|-----------------------------|------|------|--------|------|------|
|                            | -12541.1 | -1233.5 | 321.4 | 1469.2 | 10341.9 |

| Coefficients:               | Estimate | Std. Error | t value | Pr(>|t|) |
|-----------------------------|----------|------------|---------|---------|
| (Intercept)                 | 7.826e+04 | 9.961e+02  | 78.568  | < 2e-16 *** |
| GDP                         | 2.222e+00 | 7.922e-03  | 280.475 | < 2e-16 *** |
| FDI                         | 3.422e-02 | 1.036e-02  | 3.301   | 0.00103 ** |
| EM                          | -1.749e+00 | 2.357e-02  | -74.215 | < 2e-16 *** |

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Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 2596 on 536 degrees of freedom
Multiple R-squared: 0.9939, Adjusted R-squared: 0.9939
F-statistic: 2.93e+04 on 3 and 536 DF, p-value: < 2.2e-16

Source: authors

The variable AW, which turned out to be statistically insignificant, was initially removed from the “Primary Model of LP”. Testing showed that this regression model was also burdened by heteroskedasticity and autocorrelation of variables (Table 3). The elimination of undesirable phenomena from the original model required the replacement of the original variables by their differences in the “Secondary Model of LP”.

Tab. 4. Final estimation of regression coefficients in Secondary Model of LP

| Residuals:                  | Min  | 1Q   | Median | 3Q   | Max  |
|-----------------------------|------|------|--------|------|------|
|                            | -27241.9 | -717.1 | -85.6 | 872.5 | 27011.4 |

| Coefficients:               | Estimate | Std. Error | t value | Pr(>|t|) |
|-----------------------------|----------|------------|---------|---------|
| (Intercept)                 | 11.13779 | 152.80146  | 0.073   | 0.9419  |
| diff(FDI)                   | 0.02624  | 0.01084    | 2.421   | 0.0158 * |

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Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 3547 on 536 degrees of freedom
Multiple R-squared: 0.9939, Adjusted R-squared: 0.9939
F-statistic: 5485 on 2 and 536 DF, p-value: < 2.2e-16

Source: authors

By the step-by-step testing of the “Secondary Model of LP” the exogenous variables were removed, which according to the achieved t-values, showed statistical insignificance. According to the model, the variable AW is statistically insignificant. In the following modifications of the model, it was also abstracted from other variables that showed mutual multicollinearity. The final estimation of the regression coefficients at the level of significance α = 0.05 shows that if the inflow of foreign investment in OECD countries increases by one unit (dollar), labour productivity will increase by 0.02624 units (Table 4). Again, analogous as in the case of the “Secondary Model of FDI”, this result cannot be quantified entirely in this way, but it shows that there is also this opposite link between FDI flow and labour productivity compared to the previous results. That is also an important result that needs to be taken into account in decision-making processes. Based on the findings, an increase in FDI inflows could translate into an increase in labour productivity.

In the next step, the results of the cluster analysis will be presented. Hierarchical methods are based on the sequentially joining of clusters; their number decreases continuously until finally all clusters are combined into one. Wards method involves an agglomerative clustering algorithm. It looks for groups of leaves that form into branches, the branches into limbs and eventually into the trunk. Ward’s method starts out with n clusters of size 1 and continues until all the observations are included in one cluster. There were j objects in the analysed group, namely 36 OECD countries, in which were pursued k quantitative characters (2 variables – FDI and LP).
In the dendrogram, we can identify three groups of countries with similar characteristics. The groups are highlighted. There are two larger clusters with 16 or 18 countries. Outside of these clusters remained only Ireland and Luxembourg (Figure 1). To draw conclusions that consider the exact distances, we have also used a non-hierarchical clustering method, which is a scatterplot.

If we consider two variables (components), clusters can be visualised by using the non-hierarchical method K-means. Based on the previous hierarchical method, it is considered a similar number of clusters. K-means clustering is the most popular partitioning method. It requires the analyst to specify the number of clusters to extract. There are two components which explain 100 % of the point variability.

Due to testing, we have finally chosen 4 clusters as an imputed command for K-means clustering. We consider the data set, which contains 36 objects, and partition it into $k = 4$ clusters. The ellipses are based on the average and the covariance matrix of each cluster, and their size is such that they contain all the points of their cluster. The
ellipses sizes of clusters 2 and 3 are very similar. Cluster no. 3 displays more variability of Component 1. The larger degree of variability indicates the largest density of divided objects in the ellipse (Figure 2).

Clusters 2 and 3 show approximately the same variability of Component 2 (average FDI inflow). Cluster number 3 is characterised by a higher average level of both Component 2 and Component 1 (average labour productivity). Another interesting result from the regional perspective of the country of origin of the authors of this study is that all the V4 countries are situated in cluster number 2, which is characterised by low variability in average labour productivity and high variability in the average inflow of foreign investment. Outside the clusters, there are two countries, Ireland and Luxembourg, which have been separated on the basis of monitoring their mutual distance in the scatter plot (the final order was finally 4 clusters for non-hierarchical clustering).

By standardising the values of the variables of the examined objects (standardisation of FDI and LP on the scale from -1 to 1), the difference of scales of variables between the countries was eliminated in the cluster analysis. This was done because variables with a greater degree of variability have a greater impact on the resulting degree of similarity. The following Table 5 shows the selected statistics on graphically represented clusters in the scatterplot.

| Members of Cluster no. 2 (16 cases) | Mean(_FDI_) = -0.755167 | STD(_FDI_) = 0.308694 | Members of Cluster no. 3 (18 cases) | Mean(_FDI_) = -0.359110 | STD(_FDI_) = 0.067042 |
|----------------------------------|--------------------------|------------------------|----------------------------------|--------------------------|------------------------|
| Chile                            | 0.329180                |                        | Australia                        | 0.112971                |                        |
| Czech Republic                   | 0.120999                |                        | Austria                          | 0.132875                |                        |
| Estonia                          | 0.160501                |                        | Belgium                          | 0.341244                |                        |
| Greece                           | 0.357868                |                        | Canada                           | 0.212411                |                        |
| Hungary                          | 0.031410                |                        | Denmark                          | 0.166381                |                        |
| Japan                            | 0.313581                |                        | Finland                          | 0.133501                |                        |
| Korea                            | 0.182265                |                        | France                           | 0.190870                |                        |
| Latvia                           | 0.254653                |                        | Germany                          | 0.208557                |                        |
| Lithuania                        | 0.099440                |                        | Iceland                          | 0.447091                |                        |
| Mexico                           | 0.447491                |                        | Israel                           | 0.361206                |                        |
| New Zealand                      | 0.159146                |                        | Italy                            | 0.203389                |                        |
| Poland                           | 0.089628                |                        | Netherlands                      | 0.289033                |                        |
| Portugal                         | 0.047284                |                        | Norway                           | 0.557887                |                        |
| Slovak Republic                  | 0.035094                |                        | Spain                            | 0.286717                |                        |
| Slovenia                         | 0.140835                |                        | Sweden                           | 0.064818                |                        |
| Turkey                           | 0.085915                |                        | Switzerland                      | 0.451001                |                        |
|                                  |                         |                        | United Kingdom                   | 0.263845                |                        |
|                                  |                         |                        | United States                    | 0.436828                |                        |

| Members of Cluster no. 4         | Mean(_FDI_) = 1.328820  | STD(_FDI_) = 5.117805  | Members of Cluster no. 1         | Mean(_FDI_) = 4.270842  | STD(_FDI_) = 12.09999  |
|----------------------------------|--------------------------|------------------------|----------------------------------|--------------------------|------------------------|
| Ireland                          | 0.00                     |                        | Luxembourg                       | 0.00                     |                        |

The significance of each variable at the significance level α = 0.05 was confirmed by one-way analysis of variance (p-value = 0.00000). The between-cluster variability (SSB) of the LP variable is 31.465. In the case of FDI it is 33.769. The within-cluster variability (SSW) of the LP is 3.534. In the case of FDI it is 1.230.

In each of the four clusters in the table, the “centre points” of the individual variables are presented as a first, represented by the mean values (averages) of the variables and the standard deviations of the variables for the characteristics of the individual clusters. Subsequently, the countries belonging to the individual clusters and their distances from the centre point of the cluster are listed. The largest distance from the centre point in cluster no. 2 has Mexico and Chile, in cluster no. 3 it is Norway and Iceland. Significant distances of these countries from the centre points of the cluster can be clearly seen in the previous graphic presentation in the scatter plot (Figure 2), where these countries are located at the edges of the clusters and thus increase the size of the depicted ellipses. At the same time, they increase the variability of the cluster in terms of the examined variables (components). The
centre points (averages) in the case of Ireland and Luxembourg differ significantly from the centre points of larger clusters; thus, separate clusters were created represented by their position in the scatterplot.

**Discussion**

The aim of the guest country’s foreign direct investment in the host country is to increase productivity through technology transfer. On the one hand, it can have positive external effects through business relationships on the other hand (e.g., market access, improving financing conditions) and through them can be boosted economic growth. In general, two main groups of theories can be used to explain why firms engage in foreign direct investment in other countries (Firth, 1980). The first deals with attempts to maximise revenue and asset growth (Shleifer & Vishny, 1989, Morck 1990, Brandenburger & Polak, 1996). The second general group of theories focuses on the motivations for maximising value. These motivations can be financial (cash transfer), or they can be linked to synergies related to economies of scale, to efficiency gains, to gain a monopoly position (Stigler 1964, Capron & Mitchell 1998) and to efforts to strengthen managerial control over financial resources (Jensen 1988).

Boghean & State (2015) argue that the effects of foreign direct investment (FDI) on the economies of host countries are mainly related to increasing labour productivity through technology transfer, management and marketing capabilities that enable long-term technological progress and economic growth.

Investors look for countries for their investments in which they see the maximisation of benefits for their activities. On the example of V4 countries (Czech Republic, Hungary, Poland, Slovakia), which are also part of this study. Many studies from various disciplines analyse V4 market (Kolovkova & Ključnikov 2021, Žufan et al. 2020, Virglerova et al. 2020, Durda & Ključnikov 2019, Ključnikov et al. 2019). But this paper differs from these studies because it focuses on foreign direct investment inflows and the selected macroeconomic indicators in OECD countries, including V4 countries. It can be seen that after their accession to the European Union, differences in investment incentives (state aid) between member countries have decreased because of the regulations, regional competition in investment incentives has weakened. However, there is still a struggle between countries about FDI. SMEs in those countries have also been negatively affected by this fact due to facing financial problems (Civelek et al. 2020, Civelek et al. 2021, Ključnikov et al. 2021).

The first interesting result of this study is that the results of the analysis show that it is possible to talk about the link between labour productivity and the inflow of foreign direct investment in the direction from labour productivity to FDI inflow, which is in line with various findings of the authors mentioned in this study. Labour productivity in the context of this contribution contains two macroeconomic indicators (employment and GDP). It must be borne in mind that also these two macroeconomic indicators can be linked. Among economists, the “rule of thumb” was used in the pre-crisis period (before 2009) that at least 3% of GDP growth was needed for employment growth. The impact of the global economic crisis on maintaining employment showed that it would be enough 1.5% of GDP growth instead of 3%. However, employment can be negatively affected by growing global insecurity and wages in a short-term perspective. The results of this study show that the growth of the inflow of foreign direct investment can also contribute to a higher level of labour productivity, and thus there can be a larger share of the country’s GDP per person employed. This may be because if the investor is looking for opportunities to place his investment plans, he will prefer countries where the employed person is expected to produce higher added value than in the case of another investment alternative.

In view of the above findings, the article also examined the possible link between macroeconomic indicators, to which the FDI inflow indicator was added, and between the labour productivity itself (as a dependent variable). In this regard, a possible link between FDI inflow and labour productivity has been demonstrated, where it appears that even higher FDI inflow can contribute to higher labour productivity in OECD countries, which on the one hand is not in line with the findings of some of the mentioned authors, but on the other, it supports the findings of the authors as, e.g. Piscitello and Rabbiioso (2005). According to them, the impact of foreign direct investments that are realised, for example, through the acquisition of target companies, can be measured by labour productivity. Relying on the idea that multinational companies act as a means of transferring assets, which makes subsidiaries outperform the performance of their competitors in the host countries. The results of Piscitello and Rabbiioso (2005) show that foreign direct investment generally increases the labour productivity of local target companies in the medium term. Their empirical evidence concerned foreign acquisitions that took place in Italy between 1994 and 1997. The findings of this research may also result from the fact that the inflow of foreign direct investment can create new jobs on the one hand, but at the same time, it can boost and streamline the labour market, and on the other hand, it can contribute to some extent to GDP growth itself. Přívara (2021) and Přívara et al. (2020) have also offered interesting insights into the labour market. The IBM-PLI Global Location Trends 2019 study also talks about job creation. When looking at the countries of V4, it can be seen that in terms of the number of jobs created through FDI, Hungary ranked 16th in the world in 2018, while on the basis of data per million inhabitants, it ranked 5th, the Czech Republic 14th, Slovakia 15th and Poland in 19th place (IBM Institute for Business Value 2019). Apart from Hungary and Poland, the placement of the other two V4 countries deteriorated compared to the period of 2013-2017, in which Hungary was placed 5th and Poland 25th, the Czech Republic 9th, and Slovakia 6th.
Of the almost two hundred countries evaluated in the world, only Belgium, Luxembourg, Sweden, and Denmark are ahead of Hungary.

Conclusions and implications for the mining industry

The importance of foreign direct investment in the mining industry can be illustrated in the case of Guinea, whose economy relies heavily on the mining sector, as this attracts major foreign investment. As it can be seen from the research of Chin (2016), FDI improves efficiency in the mining sector. She concluded that if Guinea wants to increase the mining production, it is imperative to develop effective FDI policies. The research within this study showed that there might be a link between the FDI inflow and the labour productivity in both directions, this is why also the mining industry should count on the possibility that attracting or increasing the foreign direct investments should affect the labour productivity in a positive way (in addition to other mentioned effects or advantages).

Vrbka and Rowland (2019) have in the past dealt with the financial health of companies operating in the mining industry in the Czech Republic. Resulting from the cluster analysis, this recommendation should be applied mainly in the countries included in cluster 2, where is a relatively low level of labour productivity and there is still space to increase the FDI inflow, these countries are: Chile, Czech Republic, Estonia, Greece, Hungary, Japan, Korea, Latvia, Lithuania, Mexico, New Zealand, Poland, Portugal, Slovak Republic, Slovenia, and Turkey. Considering this point of view, from these countries, the higher potential can be seen in Mexico, Chile in the first line and in Latvia and Estonia in the second. The recommendations resulting from this study and the effects of FDI on the mining industry can also be supported by the research results of Sun and Anwar (2019), who concluded that foreign investment policy in China needs to be adjusted to attract more FDI to China’s iron ore mining industry. From the environmental point of view and the ecology, they highlighted that FDI could help develop mining technologies that are not only more efficient but also create less pollution.

The main risk factors for FDI in the mining sector can be divided into geological, political, regulatory, marketing, fiscal, monetary, environmental and social, operational, and profit (Otto, 1992; Amoatey et al., 2017; Bednarova et al., 2020). Kasych, Rowland and Yakovenko (2019), for example, discussed management tools to support the economic sustainability of companies operating in the mining industry. Other studies also provided management inspiration (Amoah et al., 2021; Bednarova et al., 2009; Belas et al., 2020; Dvorsky et al., 2021; Durana et al., 2021; Halaskova et al., 2021; Ik & Azeez, 2020; Jurkasova et al., 2016; Mura, 2021; Siwiec et al., 2019). The recommendation for the policymakers in this context is to focus on political and regulatory factors to support the inflow of FDI in the individual countries and thus to help increase the efficiency of the mining sector. Similar insights can be transferred to other areas, which favours this study (Gavurova et al., 2020; Gavurova et al., 2021; Koľveková et al., 2019; Pimonenko et al., 2021; Stefko et al., 2020; Ivankova et al., 2021; Pívrava et al., 2019). In recent years, some researchers have considered the infrastructure of a host country as a risk factor that affects foreign investment in the mining sector (Tole & Koop 2011), which could be another important task to deal with in the FDI host countries.

Based on the research performed in this study and on the concluding remarks, the future research recommendation becomes more clear. The future research recommendation proceeds from the main limitations, as the use of macroeconomic indicators for the OECD countries that are not sector-specific, so their results can be applied for the mining sector only with some suspicion. There is a need to collect data, especially on the FDI inflow in the mining sector, but there will also be a need to change the macroeconomic perspective of the rest of the examined variables with a focus on labour productivity. An interesting thing would be to include additional indicators representing the efficiency of the mining sector performance. The research should also be extended to other countries.

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