# Analysis of Investment Efficiency of Distribution Trade of Selective Countries of the European Union and Serbia on the Basis of Trust Method

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## Abstract

Investments are among the important factors in the business success of distribution trade. Given this, it is necessary to continuously investigate the factors of investment efficiency of distribution trade. The obtained results of the empirical research of the investment efficiency of the distribution trade of the selective countries of the European Union and Serbia show that Germany is in the first place. Next are France, Italy, Serbia, Bulgaria, Croatia, Hungary, Romania, Slovenia and Austria. In terms of investment efficiency, Serbia's distribution trade has taken a good position. In that respect, it is better in relation to the distribution trade between Croatia and Slovenia. This positioning of Serbia's distribution trade in terms of investment efficiency was influenced by the efficiency of human resources management, investments (investments of domestic and foreign retail chains), sales, margins and profits. Effective control of these and other relevant factors can achieve the target investment efficiency of distribution trade in Serbia.

Keywords: efficiency, factors, TRUST method, EU-Serbia distribution trade

Jel classification : L81, M31, M41, O32 DOI: 10.24818/RMCI.2022.3.382

## 1. Introduction

It is very challenging to research the investment efficiency of distribution trade. With this in mind, the subject of research in this paper is the analysis of investment efficiency of distribution trade of selective countries of the European Union (Bulgaria, Germany, France, Croatia, Italy, Hungary, Austria, Romania and Slovenia) and Serbia. The goal and purpose of this is to look at the situation in terms of investment efficiency of distribution trade of the respective countries in the function of improvement in the future by taking relevant measures.

As for the literature dedicated to the development and application of multicriteria decision-making methods, it is very rich. Also, the literature is increasingly rich and dedicated to the specifics of the application of different methods of multi-criteria decision-making in evaluating the efficiency of distributive trade (Ersoy, 2017; Gaur, 2020; Görçün, 2022; Lukic, 2020, 2021a, b). But, regardless of that, when it comes to the TRUST method, it is, as far as we know, of

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a more recent date (Torkayesh et al., 2021). Given its specifics in relation to other methods of multicriteria decision-making, especially in terms of solving complex decision-making problems, in this paper it is used in evaluating the investment efficiency of distribution trade of selective countries of the European Union and Serbia.

The basic hypothesis of the research of the given problem in this paper is that knowing the most realistic situation is a prerequisite for improving the investment efficiency of distribution trade. In order to improve, it is necessary to continuously research and effectively control the factors of investment efficiency of distribution trade. The application of the TRUST method plays a significant role in this. The necessary empirical data for the research of the given problem in this paper were collected from Eurostat.

#### 2. TRUST method

In relation to other methods of multicriteria decision-making, the TRUST (mulTi-noRmalization mUlti-distance aSsessmenT) method is specific. It is based on multi-normalization and multi-distance in solving complex decision- making problems. The multi-normalization techniques used are linear normalization based on ratio, normalization based on linear sum, linear normalization max-min and logarithmic normalization. Based on them, an aggregated normalized decision matrix is determined. Applying the TRUST method in multicriteria decision-making greatly increases the reliability (i.e., alleviates subjectivism) of normalization based on constraints (alternatives and in relation to criterion j) (Torkayesh, 2021). Euclidean, Manhattan, Lorentzian, and Pearson distance measures are used to determine the distance values of alternatives to the negative-ideal solution (Torkayesh, 2021). The TRUST method takes place procedurally through the following steps (Torkayesh, 2021):

Step 1: Determine the initial matrix. In this matrix,  $x_{ii}$  represents the initial performances of each alternative to each criterion.

$$X = [x_{ij}]_{n \times m} = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1m} \\ x_{21} & x_{22} & \cdots & x_{2m} \\ \vdots & \vdots & \vdots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{nm} \end{bmatrix}$$
(1)

Step 2: TRUST uses four normalization techniques to transform the initial decision matrix into a normalized decision matrix.

Step 2.1: Type-1 normalization attempts to calculate the normalized values of the initial decision based on a linear relationship as in the following equations (Keshavarz Ghorabaee, 2016):

$$x_{ij}^a = \frac{X_{ij}}{\max_i X_{ij}} \ IF \ j \ \in B \ (2)$$

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$$x_{ij}^{a} = \frac{\min_{i} X_{ij}}{X_{ij}} \quad IF \ j \in C \quad (3)$$

where it  $x_{ij}^a$  represents the elements of the normalized decision matrix in relation to the normalization of type - 1, B denotes the benefit and C the cost criterion, respectively.

*Step 2.2*: Normalization of type - 2 using a linear sum-based technique to normalize the elements of the initial decision matrix as in the following equations (Wen, 2020):

$$x_{ij}^{b} = \frac{X_{ij}}{\sum_{i=1}^{m} X_{ij}} \quad IF \ j \in B \quad (4)$$
$$x_{ij}^{b} = \frac{\frac{1}{x_{ij}}}{\sum_{i=1}^{m} \frac{1}{x_{ij}}} \quad IF \ j \in C \quad (5)$$

where it  $x_{ij}^{b}$  represents the elements of the normalized decision matrix in relation to the normalization of type - 2, B denotes the benefit and C the cost criterion, respectively.

*Step 2.3*: Normalization of type - 3 using the linear-min linear technique to normalize the elements of the initial decision matrix as in the following equations (Yazdani, 2019):

$$x_{ij}^{c} = \frac{\left(X_{ij} - \min_{i} X_{ij}\right)}{\left(\max_{i} X_{ij} - \min_{i} X_{ij}\right)} \ IF \ j \ \in B \quad (6)$$
$$x_{ij}^{c} = \frac{\left(\max_{i} X_{ij} - X_{ij}\right)}{\left(\max_{i} X_{ij} - \min_{i} X_{ij}\right)} \ IF \ j \ \in C \quad (7)$$

where it  $x_{ij}^c$  represents the elements of the normalized decision matrix in relation to the normalization of type - 3, B denotes the benefit and C the cost criterion, respectively.

*Step 2.4*: Finally, type - 4 normalization uses the logarithmic technique to normalize the elements of the initial decision matrix as in the following equations (Zolfani, 2020)

$$x_{ij}^d = \frac{\log(X_{ij})}{\log(\prod_{i=1}^m X_{ij})} \quad (8)$$

where it  $x_{ij}^d$  represents the elements of the normalized decision matrix in relation to the normalization of type - 4.

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Step 2.5: All normalized matrices were aggregated using four parameters  $a_i$  to construct an aggregated normalized decision matrix. Aggregation is performed using the following equation

$$h_{ij} = a_1 X_{ij}^a + a_2 X_{ij}^b + a_3 X_{ij}^c + a_4 X_{ij}^d \quad (9)$$

where the sum of all values  $a_i$  must be equal to one as in the following equation

$$a_1 + a_2 + a_3 + a_4 = 1 \quad (10)$$

Step 3: Based on preliminary constraints, the initial decision matrix is used to generate the satisfaction degree matrix (Abdeli et al., 2019, 2020). The satisfaction degree matrix is derived based on the limit values of the criteria. The elements of the satisfaction degree matrix are shown with  $f_{ij}$ . Constraint-based normalization procedure procedures are presented as follows.

Based on expert opinion, or on real data in the matrix X,  $X_j^{min} = Min(X_{ij})$  and  $X_j^{max} = Max(X_{ij})$  represent the minimum and maximum values.

Step 3.1: The constraint values for each criterion can be determined by decision makers based on their experience or the technical characteristics of each of the criteria. The constraint value is represented as an interval number  $[LB_j, UB_j]$ , where it  $LB_j$  indicates the lower limit of the limit value of criterion *j*, and  $UB_j$  the upper limit of the limit value of criterion *j*. The constraint values must be within  $x_j^{min}$  and  $x_j^{max}$  the values as in the following equation

$$Co_j = (LB_j, UB_j) \subseteq [x_j^{min}, x_j^{max}]$$
 (11)

Step 3.2: Calculate the degree of satisfaction of the initial decision matrix based on constraints. The second version of the initial decision matrix is considered as F. In this matrix,  $F_{ij}$  it represents the degree of satisfaction of the constraint of the alternative i in relation to the limited value of criterion j. The elements of the matrix F can be determined as in the following equation

For benefit criteria

$$f_{ij} = 1, if X_{ij} \in [LB_j, UB_j] \quad (12)$$

$$F_{ij} = 1 - \frac{LB_{ij} - X_{ij}}{Max \left( LB_j - X_j^{min}, X_j^{max} - UB_j \right) + 1}, if \ X_{ij} \in \left[ X_j^{min}, UB_j \right]$$
(13)

$$f_{ij} = 1 - \frac{1 - UB_j + X_{ij}}{Max \left( LB_j - X_j^{min}, X_j^{max} - UB_j \right) + 1}, if X_{ij} \in \left[ UB_j, X_j^{max} \right]$$
(14)

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For cost criteria

$$f_{ij} = \frac{1}{Max\left(LB_j - X_j^{min}, X_j^{max} - UB_j\right) + 1}, if X_{ij} \in \left[LB_j, UB_j\right]$$
(15)

$$f_{ij} = \frac{LB_j - X_{ij}}{Max \left( LB_j - X_j^{min}, X_j^{max} - UB_j \right)}, if X_{ij} \in \left[ X_j^{min}, LB_j \right]$$
(16)

$$f_{ij} = \frac{X_{ij} - UB_1}{Max \left( LB_j - X_j^{min}, X_j^{max} - UB_j \right)}, if X_{ij} \in \left[ UB_j, X_j^{max} \right]$$
(17)

Step 4: The aggregated normalized decision matrix and the constraint satisfaction degree matrix were used to generate the constrained aggregate normalized decision matrix as in the following equation. Elements of the constrained aggregate normalized decision matrix  $Y = [y_{ij}]_{n*m}$  are shown with  $y_{ij}$ .

$$y_{ij} = f_{ij}h_{ij} \quad (18)$$

Step 5: The constrained aggregate normalized decision matrix is multiplied by the weight vector included in the criteria to generate a weight constrained aggregate normalized matrix  $G = [g_{ij}]_{n*m}$  as in the following equation

$$g_{ij} = y_{ij} w_j \quad (19)$$

where it  $g_{ij}$  represents the elements of the weighted matrix, and the  $w_j$  weight of the criterion *j*.

Step 6: Determination of the negative-ideal solution of the weighted matrix *G* according to the following equation

$$\mathbb{P}_j = \min_i g_{ij} \quad (20)$$

where it  $\mathbb{Z}_j$  means a negative-ideal solution of criterion *j*.

*Step 7*: TRUST applies a two-step operation to calculate the distance of the alternative from the negative-ideal solution.

Step 7.1: In Phase 1, Euclidean and Manhattan distance measures are used as in the following equations

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$$E_i = \sqrt{\sum_{i=1}^m (g_{ij} - \mathbb{Z}_j)^2} \quad 21)$$
$$T_i = \sum_{i=1}^m |g_{ij} - \mathbb{Z}_j| \quad 22)$$

*Step 7.2*: In phase 2, the Lorentzian distance measure and the Pearson distance measure are used to calculate the distance of the alternative from the negative-ideal solution as in the following equations (Cha, 2008)

$$L_{i} = \sum_{i=1}^{m} \log(1 + |g_{ij} - \mathbb{Z}_{j}|) \quad (23)$$
$$P_{i} = \sum_{i=1}^{m} \frac{(g_{ij} - \mathbb{Z}_{j})^{2}}{\mathbb{Z}_{j}} \quad (24)$$

Step 8: According to the distance measures, the two relative estimates of the distance matrix, as in the following equations, were constructed as  $ET = |\theta_{ij}|_{n*n}$  well  $LP = |\varphi_{ij}|_{n*n}$ , respectively.

$$\theta_{ik} = (E_i - E_k) + ((E_i - E_k) * (T_i - T_k))$$
(25)  
$$\varphi_{ik} = (L_i - L_k) + ((L_i - L_k) * (P_i - P_k))$$
(26)

where  $k \in \{1, 2, ..., n\}$ .

Step 9: The combined result for each alternative,  $\Omega_i$ , is calculated as in the following equation

$$\Omega_{i} = \beta \sum_{k=1}^{n} \theta_{ik} + (1 - \beta) \sum_{k=1}^{n} \varphi_{ik} \quad (27)$$

where it  $\beta$  represents the parameter used to calculate the combined distance estimate using two relative matrix distance estimates.  $\beta$  is a negative parameter less than 1 that usually takes a value of 0.5 in decision problems. The alternatives are ranked based on their combined distance score  $\Omega_i$ , in descending order, so that the alternative with the highest value  $\Omega_i$  is considered the best alternative.

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## 3. Results and discussion

In this paper, for the purposes of measuring the investment efficiency of distribution trade of selective countries of the European Union and Serbia, the following criteria were used: C1 – persons employed - number, C2 - gross investment in tangible goods, C3 - investment per person employed, C4 - investment rate, C5 - turnover, C6 - gross margin and C7 - value added at factor cost. These criteria were chosen because they measure well the investment efficiency of distributive trade and its impact on financial performance. Alternative countries were observed: A1 - Bulgaria, A2 - Germany, A3 - France, A4 - Croatia, A5 - Italy, A6 - Hungary, A7 - Austria, A8 - Romania, A9 - Slovenia and A10 - Serbia. The selection of European Union countries was made according to which countries are the most developed and closer to the Serbian environment. Table 1 shows the initial data for 2019.

	Persons employeed - number	Gross investment in tangible goods (million euro)	Investmen t per person employed (thousand s euro)	Investme rate (investme / value added a factor cos - percenta	ent ent it st) ge	Turnover (million euro)	Gross margin (million euro)	Value added at factor cost (million euro)
Bulgaria	517483	1241.1	2.4	18.0	686	552.6	9913.3	6886.9
Germany	6586606	33021.6	5.0	9.9	212	20631.8	513781.9	334928.6
France	3364306	25085.4	7.5	13.0	138	33306.8	303984.9	192661.8
Croatia	243616	728.0	3.0	12.1	385	540.3	8223.2	6002.3
Italy	3418330	12946.6	3.8	8.9	100	)3893.9	208319.8	145338.7
Hungary	592554	2195.1	3.7	16.6	106	5833.8	20743.5	13206.6
Austria	685256	4900.1	7.2	12.5	253	3998.2	58965.0	39256.3
Romania	914741	4264.1	4.7	23.1	128	3519.6	26081.5	18488.4
Slovenia	122344	699.3	5.7	14.6	371	113.5	6460.1	4790.2
Serbia	267810	562.8	2.1	14.4	358	858.9	4886.9	3906.9
Statistics								
Mean	1671304.6000	8564.4100	4.5100	14.310 0	51	7734.9400	116136.0100	76546.6700
Median	638905.0000	3229.6000	4.2500	13.700 0	11	7676.7000	23412.5000	15847.5000
Std.	2121990.6160	11564.8639	1.87288	4.1474	73	3275.2141	173114.9230	112548.3137
Deviatio n	0	1		1	0		0	0
Minimu m	122344.00	562.80	2.10	8.90	35	858.90	4886.90	3906.90
Maximu m	6586606.00	33021.60	7.50	23.10	21	20631.80	513781.90	334928.60

Table 1. Initial data

*Note:* Statistics were calculated using SPSS software

Source: Eurostat

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	Correlations											
		1	2	3	4	5	6	7				
1 Persons employeed -	Pearson Correlation	1	.963**	.292	535	.986**	.987**	.991**				
number	Sig. (2-tailed)	10	.000	.413	.111	.000	.000	.000				
	N	10	10	10	10	10	10	10				
2 Gross investment	Pearson Correlation	.963**	1	.462	490	.990**	.989**	.985**				
in tangible	Sig. (2-tailed)	.000		.179	.150	.000	.000	.000				
goous	Ν	10	10	10	10	10	10	10				
3 Investment	Pearson Correlation	.292	.462	1	166	.391	.385	.373				
per person	Sig. (2-tailed)	.413	.179		.647	.263	.273	.288				
empioyed	N	10	10	10	10	10	10	10				
4 Investment	Pearson Correlation	535	490	166	1	566	557	567				
rate	Sig. (2-tailed)	.111	.150	.647		.088	.094	.088				
	N	10	10	10	10	10	10	10				
5 Turnover	Pearson Correlation	.986**	.990**	.391	566	1	.998**	.998**				
	Sig. (2-tailed)	.000	.000	.263	.088		.000	.000				
	N	10	10	10	10	10	10	10				
6 Gross margin	Pearson Correlation	.987**	.989**	.385	557	.998**	1	.999**				
	Sig. (2-tailed)	.000	.000	.273	.094	.000		.000				
	N	10	10	10	10	10	10	10				
7 Value added at	Pearson Correlation	.991**	.985**	.373	567	.998**	.999**	1				
factor cost	Sig. (2-tailed)	.000	.000	.288	.088	.000	.000					
	Ν	10	10	10	10	10	10	10				
**. Correlatio	n is significant at	the 0.01 le	evel (2-tail	ed).								

Table 2 shows the correlation matrix of the initial data.

Table 2. Correlations

Note: Statistics were calculated using SPSS software

Thus, there is a strong correlation between the number of employees and gross investment in tangible goods, on the one hand, and turnover, gross margin and value added at factor costs, on the other hand, at the level of statistical significance. This means, in other words, that efficient management of human resources and gross investment in tangible goods can achieve the target financial performance.

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Table 3 shows the Friedman test.

Table 5.	i i i cuman i csc
NPar Tests	
Friedman Test	
Ranks	
	Mean Rank
1 Persons employeed - number	7.00
2 Gross investment in tangible goods	3.00
3 Investment per person employed	1.00
4 Investment rate	2.00
5 Turnover	6.00
6 Gross margin	5.00
7 Value added at factor cost	4.00
Test Statistics <sup>a</sup>	
N	10
Chi-Square	60.000
df	6
Asymp. Sig.	.000
a. Friedman Test	_ ·

Table 3. Friedman Test

Note: Statistics were calculated using SPSS software

The Friedman test shows that there is a significant statistical difference between the given parameters. That is, the null hypothesis that there is no significant statistical difference between them is rejected.

In the distribution trade of Germany and France, investments are higher per employee than in Romania. Investments per employee in Romania's distribution trade are higher than in Italy. Investments per employee in Romania's distribution trade are above average, which is not the case in Serbia. In the distribution trade of Serbia, investments per employee are less than in Croatia and Slovenia. They are also smaller than Romania's distribution trade. The investment rate in Romania's distribution trade is significantly higher than in all the observed countries, which in a winning way reflected on its efficiency and financial performance. The financial performance of Romania's distribution trade, measured by gross margin, is better than in Bulgaria, Croatia, Hungary, Slovenia and Serbia, but is worse than in Germany, France, Italy and Austria.

Investments per employee in the distribution trade of Germany, France and Italy are higher than in Serbia. The investment rate in Serbia's distribution trade is higher than in Germany, France and Italy. It is higher than in Croatia, but slightly lower than in Slovenia. The rate of investment in Serbia's distribution trade is slightly higher than average. All this had an appropriate effect on the efficiency and financial performance of distribution trade in Serbia. Gross margin in Serbia's distribution trade is lower than in Croatia and Slovenia.

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The parameters of *ai* used in the TRUST model are shown below:

Parameters of ai	
a1	0.25
a2	0.25
a3	0.25
a4	0.25
β	0.5

The weighting coefficients of the criteria were determined using the AHP (Analytical Hierarchical Process) method (Saaty, 2008). They are shown in Table 2.

Tables 4 - 13 show the obtained results of the analysis of investment efficiency of selective countries of the European Union and Serbia using the TRUST method. (The calculation was performed using TRUSTSoftware-Excel.)

Initial Matr	·ix															
weights of c	riteria	0.217	73	0.1	627	0.0	)90	6 0	.1456	0.124	16		0.1247		0	0.1346
kind of crite	eria	1		1		1		1		1			1		1	l
		C1		C2		C	3	0	C <b>4</b>	C5			C6		(	C <b>7</b>
A1		5174	83	124	41.1	2.4	1	1	8	6865	2.6		9913.3		6	5886.9
A2		6586	606	33	021.6	5 5		9	.9	2120	632	2	513781	.9	3	334929
A3		3364	306	25	085.4	4 7.:	5	1	3	1383	30′	7	303984	1.9	1	192662
A4		2436	16	72	8	3		1	2.1	3854	0.3		8223.2		6	5002.3
A5		3418	330	12	946.6	5 3.8	3	8	.9	1003	894	4	208319	9.8	1	145339
A6		5925	54	21	95.1	3.2	7	1	6.6	1068	33.	8	20743.	5	1	13206.6
A7		6852	56	49	00.1	7.2	2	1	2.5	2539	98.	2	58965		3	39256.3
A8		9147	41	42	64.1	4.′	7	2	3.1	1285	19.	6	26081.	5	1	18488.4
A9		1223	44	69	9.3	5.2	7	1	4.6	3711	3.5		6460.1		4	4790.2
A10		2678	10	562	2.8	2.	1	1	4.4	3585	8.9		4886.9		3	3906.9
									1		1				1	
		Criteria	a	C1	C2	C3	;	C4	C5	C6	0	27	C8	С9		C10
Constraint V	alues	LBj		min	6	6		min	6.5	5	n	nin	6.5	5		2.5
		Ubj		max	max	к 7.5	5	max	max	max	n	nax max max		x	4	
			-													
MAX	65866	06.000	33	021.60	00	7,500	2	23.100	2120	631.800	)	513	781.900	) 3	34	928.600
MIN	12234	4.000	56	2.800		2,100	8	3.900	3585	8.900		488	6,900	3	90	6,900

14.200

5400

Table 4. Initial Matrix

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32458.800

6464262.000

MAX-MIN

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508895.000

2084772.900

391

331021.700

Type-1 Normalized Matrix							
weights of criteria	0.2173	0.1627	0.0906	0.1456	0.1246	0.1247	0.1346
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	C7
A1	0.0786	0.0376	0.3200	0.7792	0.0324	0.0193	0.0206
A2	1.0000	1.0000	0.6667	0.4286	1.0000	1.0000	1.0000
A3	0.5108	0.7597	1.0000	0.5628	0.6523	0.5917	0.5752
A4	0.0370	0.0220	0.4000	0.5238	0.0182	0.0160	0.0179
A5	0.5190	0.3921	0.5067	0.3853	0.4734	0.4055	0.4339
A6	0.0900	0.0665	0.4933	0.7186	0.0504	0.0404	0.0394
A7	0.1040	0.1484	0.9600	0.5411	0.1198	0.1148	0.1172
A8	0.1389	0.1291	0.6267	1.0000	0.0606	0.0508	0.0552
A9	0.0186	0.0212	0.7600	0.6320	0.0175	0.0126	0.0143
A10	0.0407	0.0170	0.2800	0.6234	0.0169	0.0095	0.0117

 Table 5. Type - 1 Normalized Matrix

# Table 6. Type - 2 Normalized Matrix

Type-2 Normalized Matrix							
weights of criteria	0.2173	0.1627	0.0906	0.1456	0.1246	0.1247	0.1346
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	C7
A1	0.0310	0.0145	0.0532	0.1258	0.0133	0.0085	0.0090
A2	0.3941	0.3856	0.1109	0.0692	0.4096	0.4424	0.4375
A3	0.2013	0.2929	0.1663	0.0908	0.2672	0.2617	0.2517
A4	0.0146	0.0085	0.0665	0.0846	0.0074	0.0071	0.0078
A5	0.2045	0.1512	0.0843	0.0622	0.1939	0.1794	0.1899
A6	0.0355	0.0256	0.0820	0.1160	0.0206	0.0179	0.0173
A7	0.0410	0.0572	0.1596	0.0874	0.0491	0.0508	0.0513
A8	0.0547	0.0498	0.1042	0.1614	0.0248	0.0225	0.0242
A9	0.0073	0.0082	0.1264	0.1020	0.0072	0.0056	0.0063
A10	0.0160	0.0066	0.0466	0.1006	0.0069	0.0042	0.0051

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Type-3 Normalized Matrix							
weights of criteria	0.2173	0.1627	0.0906	0.1456	0.1246	0.1247	0.1346
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	C7
A1	0.0611	0.0209	0.0556	0.6408	0.0157	0.0099	0.0090
A2	1.0000	1.0000	0.5370	0.0704	1.0000	1.0000	1.0000
A3	0.5015	0.7555	1.0000	0.2887	0.6463	0.5877	0.5702
A4	0.0188	0.0051	0.1667	0.2254	0.0013	0.0066	0.0063
A5	0.5099	0.3815	0.3148	0.0000	0.4643	0.3998	0.4273
A6	0.0727	0.0503	0.2963	0.5423	0.0340	0.0312	0.0281
A7	0.0871	0.1336	0.9444	0.2535	0.1046	0.1063	0.1068
A8	0.1226	0.1140	0.4815	1.0000	0.0444	0.0416	0.0440
A9	0.0000	0.0042	0.6667	0.4014	0.0006	0.0031	0.0027
A10	0.0225	0.0000	0.0000	0.3873	0.0000	0.0000	0.0000

 Table 7. Type - 3 Normalized Matrix

# Table 8. Type - 4 Normalized Matrix

Type-4 Normalized Matrix							
weights of criteria	0.2173	0.1627	0.0906	0.1456	0.1246	0.1247	0.1346
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	C7
A1	0.0967	0.0878	0.0615	0.1101	0.0920	0.0880	0.0875
A2	0.1154	0.1282	0.1130	0.0873	0.1204	0.1257	0.1260
A3	0.1105	0.1248	0.1414	0.0977	0.1168	0.1207	0.1205
A4	0.0912	0.0812	0.0771	0.0950	0.0873	0.0862	0.0862
A5	0.1106	0.1167	0.0937	0.0833	0.1142	0.1171	0.1177
A6	0.0977	0.0948	0.0918	0.1070	0.0957	0.0950	0.0940
A7	0.0988	0.1047	0.1386	0.0962	0.1028	0.1050	0.1048
A8	0.1009	0.1030	0.1086	0.1196	0.0972	0.0972	0.0973
A9	0.0861	0.0807	0.1222	0.1021	0.0869	0.0839	0.0839
A10	0.0919	0.0780	0.0521	0.1016	0.0867	0.0812	0.0819

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Aggregated Normalized Matrix (hij)							
× ¥/	0.217	0.162	0.090	0.145	0.124	0.124	0.134
weights of criteria	3	7	6	6	6	7	6
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	C7
A 1	0.066	0.040	0.122	0.414	0.038	0.031	0.031
AI	8	2	6	0	3	4	5
	0.627	0.628	0.356	0.163	0.632	0.642	0.640
AZ	4	4	9	9	5	0	9
4.2	0.331	0.483	0.576	0.260	0.420	0.390	0.379
A3	0	2	9	0	7	5	4
	0.040	0.029	0.177	0.232	0.028	0.029	0.029
A4	4	2	6	2	5	0	6
15	0.336	0.260	0.249	0.132	0.311	0.275	0.292
AS	0	4	9	7	5	4	2
	0.074	0.059	0.240	0.371	0.050	0.046	0.044
A6	0	3	9	0	2	1	7
	0.082	0.111	0.550	0.244	0.094	0.094	0.095
A/	7	0	7	6	1	2	0
10	0.104	0.099	0.330	0.570	0.056	0.053	0.055
A8	3	0	2	3	8	0	2
4.0	0.028	0.028	0.418	0.309	0.028	0.026	0.026
АУ	0	6	8	4	1	3	8
4.10	0.042	0.025	0.094	0.303	0.027	0.023	0.024
Alu	8	4	7	2	6	7	7

Table 9. Aggregated Normalized Matrix (hij)

	Criteria	C1	C2	C3	C4	C5	C6	C7	C8	С9	C10
	LBj	122344.000	6.000	6.000	8.900	6.500	5.000	3906.900	6.500	5.000	2.500
	UBj	6586606.000	33021.600	7.500	23.100	2120632.800	513781.900	334929.600	0.000	0.000	4,000
N 2 U	MAX (LBj- Xmin, Xmax- Jbj)	0.000	0.000	3.900	0.000	0.000	0.000	0.000	6.500	5.000	2.500

# **Table 10. Satisfaction Degree Values**

Satisfaction Degree Values							
weights of criteria	0.2173	0.1627	0.0906	0.1456	0.1246	0.1247	0.1346
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	C7
A1	1.0000	1.0000	0.2653	1.0000	1.0000	1.0000	1.0000
A2	1.0000	1.0000	0.7959	1.0000	1.0000	1.0000	1.0000
A3	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
A4	1.0000	1.0000	0.3878	1.0000	1.0000	1.0000	1.0000
A5	1.0000	1.0000	0.5510	1.0000	1.0000	1.0000	1.0000

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Satisfaction Degree Values							
weights of criteria	0.2173	0.1627	0.0906	0.1456	0.1246	0.1247	0.1346
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	<b>C7</b>
A6	1.0000	1.0000	0.5306	1.0000	1.0000	1.0000	1.0000
A7	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
A8	1.0000	1.0000	0.7347	1.0000	1.0000	1.0000	1.0000
A9	1.0000	1.0000	0.9388	1.0000	1.0000	1.0000	1.0000
A10	1.0000	1.0000	0.2041	1.0000	1.0000	1.0000	1.0000

# Table 11. Constrained Aggregated Normalized Matrix (yij)

Constrained Aggregated Normalized Matrix (yij)							
weights of criteria	0.2173	0.1627	0.0906	0.1456	0.1246	0.1247	0.1346
kind of criteria	1	1	1	1	1	1	1
	C1	C2	C3	C4	C5	C6	C7
A1	0.0668	0.0402	0.0325	0.4140	0.0383	0.0314	0.0315
A2	0.6274	0.6284	0.2841	0.1639	0.6325	0.6420	0.6409
A3	0.3310	0.4832	0.5769	0.2600	0.4207	0.3905	0.3794
A4	0.0404	0.0292	0.0689	0.2322	0.0285	0.0290	0.0296
A5	0.3360	0.2604	0.1377	0.1327	0.3115	0.2754	0.2922
A6	0.0740	0.0593	0.1278	0.3710	0.0502	0.0461	0.0447
A7	0.0827	0.1110	0.5507	0.2446	0.0941	0.0942	0.0950
A8	0.1043	0.0990	0.2426	0.5703	0.0568	0.0530	0.0552
A9	0.0280	0.0286	0.3932	0.3094	0.0281	0.0263	0.0268
A10	0.0428	0.0254	0.0193	0.3032	0.0276	0.0237	0.0247

# Table 12. Weighted Constrained Aggregated Normalized Matrix (gij)

Weighted Constrained Aggregated Normalized Matrix (gij)							
	C1	C2	C3	C4	C5	C6	C7
A1	0.0145	0.0065	0.0029	0.0603	0.0048	0.0039	0.0042
A2	0.1363	0.1022	0.0257	0.0239	0.0788	0.0801	0.0863
A3	0.0719	0.0786	0.0523	0.0379	0.0524	0.0487	0.0511
A4	0.0088	0.0048	0.0062	0.0338	0.0036	0.0036	0.0040
A5	0.0730	0.0424	0.0125	0.0193	0.0388	0.0343	0.0393

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Weighted Constrained Aggregated Normalized Matrix (gij)							
A6	0.0161	0.0096	0.0116	0.0540	0.0063	0.0057	0.0060
A7	0.0180	0.0181	0.0499	0.0356	0.0117	0.0117	0.0128
A8	0.0227	0.0161	0.0220	0.0830	0.0071	0.0066	0.0074
A9	0.0061	0.0046	0.0356	0.0450	0.0035	0.0033	0.0036
A10	0.0093	0.0041	0.0018	0.0442	0.0034	0.0030	0.0033

Negative-Ideal	Criteria	C1	C2	C3	C4	C5	C6	C7
Solution	ηи	0.0061	0.0041	0.0018	0.0193	0.0034	0.0030	0.0033

## Table 13. Results

Results									
	Alternatives	Ei	Ti	Li	Pi	ET( <del>O</del> i)	LP(qi)	Ωi	Ranking
Bulgaria	A1	0.0419	0.0562	0.0240	0.1018	-0.2045	2.4752	1.1354	5
Germany	A2	0.2138	0.4923	0.2045	11.1757	1.9545	15.9245	8.9395	1
France	A3	0.1398	0.3518	0.1487	5.6202	0.8728	5.7157	3.2942	2
Croatia	A4	0.0154	0.0237	0.0102	0.0239	-0.4245	2.6702	1.1228	6
Italy	A5	0.0979	0.2186	0.0929	2.2418	0.3392	2.3114	1.3253	3
Hungary	A6	0.0381	0.0683	0.0294	0.1484	-0.2433	2.3999	1.0783	7
Austria	A7	0.0562	0.1168	0.0500	1.4809	-0.0871	1.9811	0.9470	10
Romania	A8	0.0702	0.1239	0.0528	0.5372	0.0479	2.1047	1.0763	8
Slovenia	A9	0.0425	0.0608	0.0260	0.6898	-0.2006	2.2239	1.0117	9
Serbia	A10	0.0250	0.0280	0.0120	0.0336	-0.3432	2.6438	1.1503	4

Figure 1 shows the ranking.

Based on the obtained results of the empirical research of the investment efficiency of the distribution trade of the selective countries of the European Union and Serbia, it can be concluded that the distribution trade of Germany is in the first place in terms of investment efficiency. Next are France, Italy, Serbia, Bulgaria, Croatia, Hungary, Romania, Slovenia and Austria. The most developed countries of the European Union (Germany, France, and Italy) have the best investment efficiency in distribution trade.

In terms of investment efficiency of distribution trade, Romania is ranked eighth. The investment efficiency of distribution trade in Romania is lower than in Germany, France and Italy. It is better compared to Austria and Slovenia. This was reflected in an appropriate way in her financial performance.

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In terms of investment efficiency, Serbia's distribution trade has taken a good position. In this respect, it is better in relation to the distribution trade of the surrounding countries (Bulgaria, Croatia, Hungary, Romania, Slovenia and Austria). This had a positive effect on her financial performance.

The presented positioning of distribution trade in terms of investment efficiency of the observed countries was influenced by human resource management, investment (domestic and foreign retail chains), sales management, margin and profit. Effective control of these and other relevant factors can achieve the target investment efficiency of distribution trade.

In relation to the ratio analysis, the application of the TRUST method provides a more realistic ranking of distribution trade in terms of investment efficiency by individual countries. Therefore, it is recommended.

## 4. Conclusion

Based on the conducted empirical research using statistical analysis and the TRUST method, the following can be concluded:

1. In the distribution trade of Germany and France, investments are higher per employee than in Romania. Investments per employee in Romania's distribution trade are higher than in Italy. Investments per employee in Romania's distribution trade are above average, which is not the case in Serbia. In the distribution trade of Serbia, investments per employee are less than in Croatia and Slovenia. They are also smaller than Romania's distribution trade. The investment rate in Romania's distribution trade is significantly higher than in all the observed countries, which in a winning way reflected on its efficiency and financial performance. The financial performance of Romania's distribution trade, measured by gross margin, is better

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than in Bulgaria, Croatia, Hungary, Slovenia and Serbia, but is worse than in Germany, France, Italy and Austria.

2. Investments per employee in the distribution trade of Germany, France and Italy are higher than in Serbia. The investment rate in Serbia's distribution trade is higher than in Germany, France and Italy. It is higher than in Croatia, but slightly lower than in Slovenia. The rate of investment in Serbia's distribution trade is slightly higher than average. All this had an appropriate effect on the efficiency and financial performance of distribution trade in Serbia. Gross margin in Serbia's distribution trade is lower than in Croatia and Slovenia.

3. The obtained results of the empirical research of investment efficiency of distribution trade of selective countries of the European Union and Serbia show that the distribution trade of Germany is in the first place in terms of investment efficiency. Next are France, Italy, Serbia, Bulgaria, Croatia, Hungary, Romania, Slovenia and Austria. The most developed countries of the European Union (Germany, France, and Italy) are best positioned in terms of investment efficiency and distribution trade.

4. The investment efficiency of distribution trade in Romania is lower than in Germany, France and Italy. It is better compared to Austria and Slovenia. This was reflected in an appropriate way in her financial performance.

5. In terms of investment efficiency, Serbia's distribution trade has taken a good position. In this respect, it is better in relation to the distribution trade of the surrounding countries (Bulgaria, Croatia, Hungary, Romania, Slovenia and Austria).

6. The presented positioning of distribution trade in terms of investment efficiency of the observed countries was influenced by efficient management of human resources, investment (domestic and foreign retail chains), sales, margins and profits. Effective control of these and other relevant factors can achieve the target investment efficiency in the distribution trade of the observed countries. The effect of this is to improve financial performance.

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