

Analysis of Electronic Business of European Union and Serbia Companies Based on the LMAW-DNMA Method

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Modern companies perform their business activities in the classic way and electronically. Recently, there is an increasingly pronounced trend for companies from all countries of the world, which means both the European Union and Serbia, to conduct business activities electronically. Electronic business enables the performance of business activities in real time, without geographical and time barriers. The effects of electronic business are, among other things, an increase in sales, a reduction in costs and thus an increase in profits. In this paper, the trend and rankings of the electronic business of companies in the European Union and Serbia are comparatively analyzed. The general conclusion is that the participation of electronic business in the overall business of companies in the European Union and Serbia is increasing. This is almost the case with companies all over the world. The electronic business of Serbian companies is at a lower level compared to companies in the European Union. It is also at a lower level compared to companies in Croatia and Slovenia. Considering the positive effects of electronic business, it is necessary for Serbian companies to invest as much as possible in information and communication technology.

Keywords: Website, Use of Enterprise Resource Planning (ERP), Use of Customer Relationship Management (CRM), enterprise, European Union, Serbia, LMAW-DNMA method

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1 Introduction

Companies today perform their business activities in the classic way and electronically. Research on the development of electronic business of all companies is very challenging, current, important and complex [1-10], [14-20]. Statistics record the trend of increasing electronic business in the total business of companies in all countries of the world, which means the European Union and Serbia [10-13]. This is completely understandable when you consider the fact that business activities can be carried out in real time, without time and geographical barriers. The effects of this are, among other things, an increase in sales, a reduction in costs and thus an increase in profits. It should be emphasized that in the conditions of the Covid-19 corona virus pandemic, the importance of developing and

performing, in addition to traditional business, electronic business has come to the fore.

In this paper, starting from the actuality and importance, on the basis of statistical data of Eurostat, the reached level of development and ranking of the electronic business of the companies of the European Union and Serbia is comparatively analyzed. In addition to the classic methodology, the LMAW-DNMA method is used. The LMAW-DNMA method, as a multi-criteria decision-making method, enables the selection and ranking of companies from the European Union and Serbia based on a number of criteria according to the development of electronic business. Based on that, the current situation and the need for the development of the company's electronic business in order to achieve the target performance can be realistically assessed.

application of the LMAW-DNMA method [4], [5], [9], [14].

The LMAW (Logarithm Methodology of Additive Weights) method is the latest method used to calculate criteria weights and rank

2 Research methodology

The research methodology of the treated problem in this paper is based on the classical methodology and, in particular, on the

alternatives [9], [4]. It takes place through the following steps : m alternatives $A = \{A_1, A_2, \dots, A_m\}$ are evaluated in comparison with n criteria $C = \{C_1, C_2, \dots, C_n\}$ with the participation of k experts $E = \{E_1, E_2, \dots, E_k\}$ and according to a predefined linguistic scale [14].

Step 1: Determination of weight coefficients of criteria

Experts $E = \{E_1, E_2, \dots, E_k\}$ set priorities with criteria $C = \{C_1, C_2, \dots, C_n\}$ in relation to previously defined values of the linguistic scale. At the same time, they assign a higher

$$\gamma_{AIP} = \frac{\gamma_{min}^e}{S}$$

where is γ_{min}^e the minimum value of the priority vector and S should be greater than the base logarithmic function. In the case of using the function Ln, the value of S can be chosen as 3.

$$n_{Cn}^e = \frac{\gamma_{Cn}^e}{\gamma_{AIP}} \quad (1)$$

So the relational vector $R^e = (n_{C1}^e, n_{C2}^e, \dots, n_{Cn}^e)$ is obtained. Where it n_{Cn}^e represents the value of the real vector derived from the previous equation.

:

$$w_j^e = \frac{\log_A(n_{Cn}^e)}{\log_A(\prod_{j=1}^n n_{Cn}^e)}, A > 1 \quad (2)$$

where w_j^e it represents the weighting coefficients obtained according to expert evaluations e^{th} and the n_{Cn}^e elements of the realization vector R . The obtained values for the weighting coefficients must meet the condition that $\sum_{j=1}^n w_j^e = 1$.

$$W_j = \left(\frac{1}{k \cdot (k-1)} \cdot \sum_{x=1}^k (w_j^{(x)})^p \cdot \sum_{\substack{y=1 \\ y \neq x}}^k (w_{ij}^{(y)})^q \right)^{\frac{1}{p+q}} \quad (3)$$

The value of p and q are stabilization parameters and $p, q \geq 0$. The resulting weight coefficients should fulfill the condition that $\sum_{j=1}^n w_j = 1$.

DNMA (Double Normalization-based Multiple Aggregation) method is a newer

value to the criterion of greater importance and a lower value to the criterion of less importance on the linguistic scale. By the way, the priority vector is obtained. The label γ_{Cn}^e represents the value of the linguistic scale that the expert $e (1 \leq e \leq k)$ assigns to the criterion $C_t (1 \leq t \leq n)$.

Step 1.1: Defining the absolute anti-ideal point γ_{AIP}

The absolute ideal point should be less than the smallest value in the priority vector. It is calculated according to the equation:

Step 1.2: Determining the relationship between the priority vector and the absolute anti-ideal point

The relationship between the priority vector and the absolute anti-ideal point is calculated using the following equation:

Step 1.3: Determination of the vector of weight coefficients

The vector of weight coefficients $w = (w_1, w_2, \dots, w_n)^T$ is calculated by the expert $e (1 \leq e \leq k)$ using the following equation

By applying the Bonferroni aggregator shown in the following equation, the aggregated vector of weight coefficients is determined $w = (w_1, w_2, \dots, w_n)^T$:

method for showing alternatives [4]. Two different normalized (linear and vector) techniques are used, as well as three different coupling functions (full compensation - CCM, non-compensation - UCM and incomplete

compensation - ICM). The steps of applying this method are as follows [9]:

Step 1: Normalized decision matrix

$$\hat{x}_{ij}^{1N} = 1 - \frac{|x^{ij} - r_j|}{\max \left\{ \max_i x^{ij}, r_j \right\} - \min \left\{ \min_i x^{ij}, r_j \right\}} \quad (4)$$

The vector (\hat{x}_{ij}^{2N}) is normalized using the following equation:

$$\hat{x}_{ij}^{2N} = 1 - \frac{|x^{ij} - r_j|}{\sqrt{\sum_{i=1}^m (x^{ij})^2 + (r_j)^2}} \quad (5)$$

The value r_j is the target value for c_j the criterion and is considered $\max_i x^{ij}$ for both utility and $\min_i x^{ij}$ cost criteria.

Step 2: Determining the weight of the criteria

This step consists of three phases:

$$\sigma_j = \sqrt{\frac{\sum_{i=1}^m \left(\frac{x^{ij}}{\max_i x^{ij}} - \frac{1}{m} \sum_{i=1}^m \left(\frac{x^{ij}}{\max_i x^{ij}} \right) \right)^2}{m}} \quad (6)$$

Step 2.2: Values of the standard deviation calculated for the criteria se

$$w_j^\sigma = \frac{\sigma_j}{\sum_{i=1}^n \sigma_j} \quad (7)$$

Step 2.3: Finally, the weights are adjusted with the following equation:

$$\hat{w}_j = \frac{\sqrt{w_j^\sigma \cdot w_j}}{\sum_{i=1}^n \sqrt{w_j^\sigma \cdot w_j}} \quad (8)$$

Step 3: Calculating the aggregation model Three aggregation functions (CCM, UCM and ICM) are calculated separately for each alternative.

The elements of the decision matrix are normalized with linear (\hat{x}_{ij}^{1N}) normalization using the following equation:

Step 2.1: In this phase, the standard deviation (σ_j) for the criterion c_j is determined with the following equation where m is the number of alternatives:

normalize with the following equation:

The CCM (Complete Compensation Model) is calculated using the following equation:

$$u_1(a_i) = \sum_{j=1}^n \frac{\hat{w}_j \cdot \hat{x}_{ij}^{1N}}{\max_i \hat{x}_{ij}^{1N}} \quad (9)$$

The UCM (non-compensatory model) is calculated using the following equation:

$$u_2(a_i) = \max_j \hat{w}_j \left(\frac{1 - \hat{x}_{ij}^{1N}}{\max_i \hat{x}_{ij}^{1N}} \right) \quad (10)$$

The ICM (Incomplete Compensation Model) is calculated using the following equation:

$$u_3(a_i) = \prod_{j=1}^n \left(\frac{\hat{x}_{ij}^{2N}}{\max_i \hat{x}_{ij}^{2N}} \right)^{\hat{w}_j} \quad (11)$$

Step 4: Integration of utility values with the following equation using the The calculated utility functions are integrated Euclidean principle of distance:

$$DN_i = w_1 \sqrt{\varphi \left(\frac{u_1(a_i)}{\max_i u_1(a_i)} \right)^2 + (1 - \varphi) \left(\frac{m - r_{1(a_i)+1}}{m} \right)^2} - w_2 \sqrt{\varphi \left(\frac{u_2(a_i)}{\max_i u_2(a_i)} \right)^2 + (1 - \varphi) \left(\frac{r_2(a_i)}{m} \right)^2} + w_3 \sqrt{\varphi \left(\frac{u_3(a_i)}{\max_i u_3(a_i)} \right)^2 + (1 - \varphi) \left(\frac{m - r_3(a_i) + 1}{m} \right)^2} \quad (12)$$

In this case, the means $r_1(a_i)$ and $r_3(a_i)$ represent the ordinal number of the alternative a_i sorted by CCM and ICM functions in descending value (higher value first). On the other hand, $r_2(a_i)$ it shows the sequence number in the obtained order according to the increasing value (smaller value first) for the UCM function used. The label φ is the relative importance of the child value used and is in the range [0,1]. It is considered that it can be taken as $\varphi = 0.5$. The coefficients w_1, w_2, w_3 are obtained weights of the used functions CCM, UCM and ICM, respectively. The sum should be equal $w_1 + w_2 + w_3 = 1$. When determining the weights, if the decision maker attaches importance to a wider range of performance alternatives, he can set a higher value for w_1 . In case the decision maker is not willing to take risks, i.e. to choose a poor alternative according to some criterion, he can assign a higher weight to w_2 . However, the decision maker may assign a greater weight to w_3 if he

simultaneously considers overall performance and risk. Finally, the DN values are sorted in descending order, with the higher value alternatives being the best.

3 Ranking of companies in the European Union and Serbia according to the development of electronic business using the LMAW-DNMA method

When ranking the companies of the European Union and Serbia according to the development of electronic business based on the LMAW-DNMA method, the following criteria were used: C1 - Website, C2 - Use of Enterprise Resource Planning (ERP) and C3 - Use of Customer Relationship Management (CRM). According to statistics from Eurostat, they are good indicators of electronic business. Alternatives are the member states of the European Union and Serbia. Table 1 shows the criteria, alternatives and initial data for 2021.

Table 1. Enterprises adopting e-business applications, 2021 (% enterprise)

		Website	Use of Enterprise Resource Planning (ERP)	Use of Customer Relationship Management (CRM)
		C1	C2	C3
	EU	78	38	35
A1	Belgium	:	57	54
A2	Bulgaria	52	22	17
A3	Czechia	83	38	18
A4	Denmark	:	50	42
A5	Germany	89	38	45
A6	Estonia	:	23	23
A7	Ireland	82	24	32
A8	Greece	62	35	20
A9	Spain	77	49	40
A10	France	70	45	32
A11	Croatia	68	24	20
A12	Italy	75	32	27
A13	Cyprus	69	34	39
A14	Latvia	68	39	18
A15	Lithuania	78	45	32
A16	Luxembourg	81	40	35
A17	Hungary	63	21	15
A18	Malta	82	39	39
A19	Netherlands	92	43	52
A20	Austria	91	45	46
A21	Poland	71	32	32
A22	Portugal	62	52	25
A23	Romania	51	17	17
A24	Slovenia	83	36	22
A25	Slovakia	76	31	22
A26	Finland	96	48	46
A27	Sweden	91	35	38
	Norway	83	39	39
	Albania	44	35	24
	Montenegro ⁽¹⁾	: (in)	: (in)	47
	North Macedonia	49	15	15
A28	Serbia	85	22	14
	Turkey	49	28	11
	Bosnia and Herzegovina	62	26	16

Note: 2021 EU aggregates for website variables excluding Belgium, Denmark and Estonia. (:) data not available. (¹) Montenegro: data for enterprises with website and for use of ERP: unreliable

Source: Eurostat (isoc_ciweb) and (isoc_eb_iip)

In the countries of the European Union, the trend is to increase the electronic business of companies. This is clearly shown by the data in Table 2. For illustration, in 2021

compared to 2019, there was an increase in Enterprises with a website (from 77% to 78%), Use of Enterprise Resource Planning (ERP) (from 36% to 38 %) and Use of

Customer Relationship Management (CRM) (from 33% to 35%).

Table 2. Adoption of e-business applications in enterprises, EU, 2019 and 2021 (% of enterprises)

	2019	2021
Enterprises with a website	77	78
Use of Enterprise Resource Planning (ERP)	36	38
Use of Customer Relationship Management (CRM)	33	35

Source: Eurostat (isoc_ciweb) and (isoc_eb_iip)

In Serbia, the percentage of companies that use the website is higher than in North Macedonia and Bosnia and Herzegovina. However, compared to Croatia and Slovenia, the percentage of companies in Serbia that use the website is lower. The percentage of companies in Serbia that use *Use of Enterprise Resource Planning (ERP)* is lower than in Croatia and Slovenia. Compared to North Macedonia, it is larger, but smaller compared to Bosnia and Herzegovina. In Serbia, the percentage of companies that use *Use of Customer Relationship Management (CRM)* is lower than in Croatia and Slovenia. Likewise, it is smaller compared to North

Macedonia and Bosnia and Herzegovina. All in all, the level of development of the electronic business of companies in Serbia is unsatisfactory, especially in relation to the countries of the European Union. In the future, due to its importance, it is therefore necessary that companies in Serbia perform their business activities electronically as much as possible. For these needs, they should invest more in information and communication technology.

The weight coefficients of the criteria were calculated using the LMAW method. Table 3 shows the prioritization scale for those needs.

Table 3. Prioritization Scale

Linguistic Variables	Abbreviation	Prioritization
Absolutely Low	AL	1
Very Low	VL	1.5
Low	L	2
Medium	M	2.5
Equal	E	3
Medium High	MH	3.5
High	H	4
Very High	VH	4.5
Absolutely High	AH	5

Table 4 shows the criteria evaluation coefficient. procedure and, as a result, their weighted

Table 4. Evaluation and weight coefficients of the criteria

KIND	1	1	1
	C1	C2	C3
E1	H	AH	H
E2	VH	VH	MH
E3	H	MH	VH
E4	MH	H	H

E5	VH	VH	MH
YAIP	0.5		
	C1	C2	C3
R1	8	10	8
R2	9	9	7
R3	8	7	9
R4	7	8	8
R5	9	9	7
Weight Coefficients Vector	C1	C2	C3
W1j	0.322	0.356	0.322
W2j	0.347	0.347	0.307
W3j	0.334	0.313	0.353
W4j	0.319	0.341	0.341
W5j	0.347	0.347	0.307
Aggregated Fuzzy Vectors	C1	C2	C3
W1j	0.022	0.024	0.021
W2j	0.023	0.024	0.020
W3j	0.022	0.022	0.023
W4j	0.022	0.023	0.022
W5j	0.023	0.024	0.020
SUM	0.111	0.116	0.106
Aggregated Weight Coefficient Vectors	0.3335	0.3405	0.3257

According to the LMAW method, the most important criterion is C2 - Use of Enterprise Resource Planning (ERP). By increasing the use of Use of Enterprise Resource Planning (ERP), it is possible to influence the increase

in the performance of companies in the countries of the European Union and Serbia. Tables 5 - 11 show the procedure and results of applying the LMAW-DNMA method.

Table 5. Initial Matrix

INITIAL MATRIX	KIND	1	1	1
	Weight	0.3335	0.3405	0.3257
		C1	C2	C3
A1	0	57	54	
A2	52	22	17	
A3	83	38	18	
A4	0	50	42	
A5	89	38	45	
A6	0	23	23	
A7	82	24	32	
A8	62	35	20	
A9	77	49	40	
A10	70	45	32	
A11	68	24	20	
A12	75	32	27	
A13	69	34	39	
A14	68	39	18	
A15	78	45	32	

	A16	81	40	35
	A17	63	21	15
	A18	82	39	39
	A19	92	43	52
	A20	91	45	46
	A21	71	32	32
	A22	62	52	25
	A23	51	17	17
	A24	83	36	22
	A25	76	31	22
	A26	96	48	46
	A27	91	35	38
	A28	85	22	14
	MAX	96.0000	57.0000	54.0000
	MIN	0.0000	17.0000	14.0000

Table 6. Linear Normalization Matrix

Linear Normalization MATRIX		C1	C2	C3	MAX
		A1	0.0000	1.0000	1.0000
	A2	0.5417	0.1250	0.0750	0.5417
	A3	0.8646	0.5250	0.1000	0.8646
	A4	0.0000	0.8250	0.7000	0.8250
	A5	0.9271	0.5250	0.7750	0.9271
	A6	0.0000	0.1500	0.2250	0.2250
	A7	0.8542	0.1750	0.4500	0.8542
	A8	0.6458	0.4500	0.1500	0.6458
	A9	0.8021	0.8000	0.6500	0.8021
	A10	0.7292	0.7000	0.4500	0.7292
	A11	0.7083	0.1750	0.1500	0.7083
	A12	0.7813	0.3750	0.3250	0.7813
	A13	0.7188	0.4250	0.6250	0.7188
	A14	0.7083	0.5500	0.1000	0.7083
	A15	0.8125	0.7000	0.4500	0.8125
	A16	0.8438	0.5750	0.5250	0.8438
	A17	0.6563	0.1000	0.0250	0.6563
	A18	0.8542	0.5500	0.6250	0.8542
	A19	0.9583	0.6500	0.9500	0.9583
	A20	0.9479	0.7000	0.8000	0.9479
	A21	0.7396	0.3750	0.4500	0.7396
	A22	0.6458	0.8750	0.2750	0.8750
	A23	0.5313	0.0000	0.0750	0.5313
	A24	0.8646	0.4750	0.2000	0.8646
	A25	0.7917	0.3500	0.2000	0.7917
	A26	1.0000	0.7750	0.8000	1.0000
	A27	0.9479	0.4500	0.6000	0.9479
	A28	0.8854	0.1250	0.0000	0.8854

Table 7. Vector Normalization Matrix

Vector Normalization MATRIX		C1	C2	C3	MAX
	A1	0.0000	1.0000	1.0000	1.0000
	A2	0.8889	0.8315	0.7971	0.8889
	A3	0.9672	0.9085	0.8026	0.9672
A4	0.0000	0.9663	0.9342	0.9663	
A5	0.9823	0.9085	0.9507	0.9823	
A6	0.0000	0.8363	0.8300	0.8363	
A7	0.9646	0.8411	0.8794	0.9646	
A8	0.9141	0.8941	0.8136	0.9141	
A9	0.9520	0.9615	0.9232	0.9615	
A10	0.9343	0.9422	0.8794	0.9422	
A11	0.9293	0.8411	0.8136	0.9293	
A12	0.9470	0.8797	0.8520	0.9470	
A13	0.9318	0.8893	0.9178	0.9318	
A14	0.9293	0.9133	0.8026	0.9293	
A15	0.9545	0.9422	0.8794	0.9545	
A16	0.9621	0.9182	0.8958	0.9621	
A17	0.9166	0.8267	0.7862	0.9166	
A18	0.9646	0.9133	0.9178	0.9646	
A19	0.9899	0.9326	0.9890	0.9899	
A20	0.9874	0.9422	0.9561	0.9874	
A21	0.9368	0.8797	0.8794	0.9368	
A22	0.9141	0.9759	0.8410	0.9759	
A23	0.8863	0.8074	0.7971	0.8863	
A24	0.9672	0.8989	0.8245	0.9672	
A25	0.9495	0.8748	0.8245	0.9495	
A26	1.0000	0.9567	0.9561	1.0000	
A27	0.9874	0.8941	0.9123	0.9874	
A28	0.9722	0.8315	0.7807	0.9722	
Adj Wj		0.2973	0.3406	0.3622	

Table 8. CCM (Complete Compensatory Model)

CCM (Complete Compensatory Model)	u1(ai)	C1	C2	C3	SUM
	A1	0.0000	0.3406	0.3622	0.7027
	A2	0.2973	0.0786	0.0502	0.4260
	A3	0.2973	0.2068	0.0419	0.5459
A4	0.0000	0.3406	0.3073	0.6479	
A5	0.2973	0.1929	0.3028	0.7929	
A6	0.0000	0.2270	0.3622	0.5892	
A7	0.2973	0.0698	0.1908	0.5578	
A8	0.2973	0.2373	0.0841	0.6187	
A9	0.2973	0.3397	0.2935	0.9304	
A10	0.2973	0.3269	0.2235	0.8477	
A11	0.2973	0.0841	0.0767	0.4581	
A12	0.2973	0.1635	0.1507	0.6114	
A13	0.2973	0.2014	0.3150	0.8136	
A14	0.2973	0.2644	0.0511	0.6128	
A15	0.2973	0.2934	0.2006	0.7913	

	A16	0.2973	0.2321	0.2254	0.7547
	A17	0.2973	0.0519	0.0138	0.3629
	A18	0.2973	0.2193	0.2650	0.7816
	A19	0.2973	0.2310	0.3590	0.8873
	A20	0.2973	0.2515	0.3057	0.8544
	A21	0.2973	0.1727	0.2204	0.6903
	A22	0.2194	0.3406	0.1138	0.6738
	A23	0.2973	0.0000	0.0511	0.3484
	A24	0.2973	0.1871	0.0838	0.5681
	A25	0.2973	0.1506	0.0915	0.5393
	A26	0.2973	0.2639	0.2898	0.8509
	A27	0.2973	0.1617	0.2293	0.6882
	A28	0.2973	0.0481	0.0000	0.3453

Table 9. UCM (Uncompensatory Model)

UCM (Uncompensatory Model)	u2(ai)	C1	C2	C3	MAX
		A1	0.0000	0.0000	0.0000
	A2	0.0000	0.2620	0.3120	0.3120
	A3	0.0000	0.1338	0.3203	0.3203
	A4	0.0000	0.0000	0.0549	0.0549
	A5	0.0000	0.1477	0.0594	0.1477
	A6	0.0000	0.1135	0.0000	0.1135
	A7	0.0000	0.2708	0.1714	0.2708
	A8	0.0000	0.1033	0.2781	0.2781
	A9	0.0000	0.0009	0.0687	0.0687
	A10	0.0000	0.0136	0.1387	0.1387
	A11	0.0000	0.2564	0.2855	0.2855
	A12	0.0000	0.1771	0.2115	0.2115
	A13	0.0000	0.1392	0.0472	0.1392
	A14	0.0000	0.0761	0.3111	0.3111
	A15	0.0000	0.0472	0.1616	0.1616
	A16	0.0000	0.1085	0.1368	0.1368
	A17	0.0000	0.2887	0.3484	0.3484
	A18	0.0000	0.1213	0.0972	0.1213
	A19	0.0000	0.1096	0.0031	0.1096
	A20	0.0000	0.0891	0.0565	0.0891
	A21	0.0000	0.1679	0.1418	0.1679
	A22	0.0779	0.0000	0.2484	0.2484
	A23	0.0000	0.3406	0.3111	0.3406
	A24	0.0000	0.1535	0.2784	0.2784
	A25	0.0000	0.1900	0.2707	0.2707
	A26	0.0000	0.0766	0.0724	0.0766
	A27	0.0000	0.1789	0.1329	0.1789
	A28	0.0000	0.2925	0.3622	0.3622

Table 10. ICM (Incomplete Compensatory Model)

ICM (Incomplete Compensatory Model)	u3(ai)	C1	C2	C3	MAX
	A1	0.0000	1.0000	1.0000	0.0000
A2	1.0000	0.9775	0.9613	0.9397	
A3	1.0000	0.9789	0.9347	0.9150	
A4	0.0000	1.0000	0.9878	0.0000	
A5	1.0000	0.9738	0.9882	0.9623	
A6	0.0000	1.0000	0.9973	0.0000	
A7	1.0000	0.9544	0.9670	0.9230	
A8	1.0000	0.9925	0.9587	0.9515	
A9	0.9971	1.0000	0.9854	0.9825	
A10	0.9975	1.0000	0.9753	0.9729	
A11	1.0000	0.9666	0.9530	0.9212	
A12	1.0000	0.9752	0.9624	0.9386	
A13	1.0000	0.9842	0.9945	0.9788	
A14	1.0000	0.9941	0.9483	0.9427	
A15	1.0000	0.9956	0.9707	0.9665	
A16	1.0000	0.9842	0.9745	0.9591	
A17	1.0000	0.9654	0.9459	0.9132	
A18	1.0000	0.9816	0.9821	0.9640	
A19	1.0000	0.9799	0.9997	0.9796	
A20	1.0000	0.9842	0.9884	0.9728	
A21	1.0000	0.9788	0.9773	0.9566	
A22	0.9807	1.0000	0.9475	0.9293	
A23	1.0000	0.9688	0.9623	0.9322	
A24	1.0000	0.9754	0.9439	0.9206	
A25	1.0000	0.9725	0.9502	0.9241	
A26	1.0000	0.9850	0.9839	0.9692	
A27	1.0000	0.9668	0.9718	0.9395	
A28	1.0000	0.9482	0.9236	0.8757	

Table 11. Ranking of alternatives

											w1	w2	w3	
											0.6	0.1	0.3	
		CCM		φ	UCM		φ	ICM		φ	Utility Values		Rank Order	
		u1(ai)	Rank	0.5	u2(ai)	Rank	0.5	u3(ai)	Rank	0.5				
Belgium	A1	0.7027	11	0.7013	0.0000	1	0.0253	0.0000	26	0.0758	0.4461	0.4461	26	
Bulgaria	A2	0.4260	25	0.3391	0.3120	24	0.8593	0.9397	14	0.7752	0.5220	0.5220	21	
Czechia	A3	0.5459	22	0.4510	0.3203	25	0.8886	0.9150	23	0.6757	0.5622	0.5622	19	
Denmark	A4	0.6479	15	0.6062	0.0549	2	0.1184	0.0000	26	0.0758	0.3983	0.3983	27	
Germany	A5	0.7929	7	0.8196	0.1477	12	0.4183	0.9623	9	0.8572	0.7907	0.7907	8	
Estonia	A6	0.5892	19	0.5141	0.1135	7	0.2835	0.0000	26	0.0758	0.3595	0.3595	28	
Ireland	A7	0.5578	21	0.4696	0.2708	19	0.7139	0.9230	20	0.7021	0.5638	0.5638	18	
Greece	A8	0.6187	16	0.5734	0.2781	20	0.7415	0.9515	12	0.8082	0.6607	0.6607	14	
Spain	A9	0.9304	1	1.0000	0.0687	3	0.1540	0.9825	1	1.0000	0.9154	0.9154	1	
France	A10	0.8477	5	0.8845	0.1387	10	0.3702	0.9729	4	0.9428	0.8506	0.8506	4	
Croatia	A11	0.4581	24	0.3703	0.2855	22	0.7870	0.9212	21	0.6931	0.5088	0.5088	22	
Italy	A12	0.6114	18	0.5413	0.2115	16	0.5777	0.9386	16	0.7510	0.6079	0.6079	16	
Cyprus	A13	0.8136	6	0.8483	0.1392	11	0.3886	0.9788	3	0.9630	0.8368	0.8368	5	

Latvia	A14	0.6128	17	0.5556	0.3111	23	0.8403	0.9427	13	0.7897	0.6543	0.6543	15
Lithuania	A15	0.7913	8	0.8018	0.1616	13	0.4553	0.9665	7	0.8902	0.7937	0.7937	7
Luxembourg	A16	0.7547	10	0.7478	0.1368	9	0.3507	0.9591	10	0.8406	0.7359	0.7359	10
Hungary	A17	0.3629	26	0.2860	0.3484	27	0.9631	0.9132	24	0.6693	0.4687	0.4687	24
Malta	A18	0.7816	9	0.7797	0.1213	8	0.3112	0.9640	8	0.8733	0.7609	0.7609	9
Netherlands	A19	0.8873	2	0.9590	0.1096	6	0.2621	0.9796	2	0.9808	0.8958	0.8958	2
Austria	A20	0.8544	3	0.9234	0.0891	5	0.2149	0.9728	5	0.9260	0.8534	0.8534	3
Poland	A21	0.6903	12	0.6779	0.1679	14	0.4821	0.9566	11	0.8250	0.7024	0.7024	11
Portugal	A22	0.6738	14	0.6369	0.2484	17	0.6476	0.9293	18	0.7242	0.6642	0.6642	13
Romania	A23	0.3484	27	0.2695	0.3406	26	0.9344	0.9322	17	0.7362	0.4760	0.4760	23
Slovenia	A24	0.5681	20	0.4879	0.2784	21	0.7594	0.9206	22	0.6857	0.5744	0.5744	17
Slovakia	A25	0.5393	23	0.4370	0.2707	18	0.6971	0.9241	19	0.7114	0.5453	0.5453	20
Finland	A26	0.8509	4	0.9038	0.0766	4	0.1805	0.9692	6	0.9077	0.8326	0.8326	6
Sweden	A27	0.6882	13	0.6609	0.1789	15	0.5152	0.9395	15	0.7630	0.6770	0.6770	12
Serbia	A28	0.3453	28	0.2637	0.3622	28	1.0000	0.8757	25	0.6383	0.4497	0.4497	25
	MAX	0.9304			0.3622			0.9825					

Therefore, according to the results of the LMAW-DNMA method, the top five companies of the European Union countries in terms of the development of electronic business fall in the following order: Spain, Netherlands, Austria, France and Cyprus. Companies from Germany (eighth place), France (fourth place) and Italy (sixteenth place) are positioned at an enviable level. The worst position was taken by Estonian companies. Romanian companies are positioned on the twenty-third place.

Serbian companies are positioned in twenty-fifth place. They are in relation to the companies of the European Union, and in the region, that is. Croatia (twenty-second place) and Slovenia (seventeenth place) are worse positioned. All of this in itself leads to the conclusion that companies in Serbia should,

given its importance, pay more attention to the development of e-business in the future.

5 Electronic business of European Union companies by size

In the European Union, the development of electronic business of companies varies by size. The data in Table 12 clearly supports this. The development of electronic business in large companies is significantly higher than in small and medium-sized ones. Likewise, the development of electronic business in medium-sized companies is higher than in small ones. The conclusion is that the size of the company significantly affects the level of application of electronic business. Larger companies are able to invest more in information and communication technology than smaller ones.

Table 12. Enterprises adopting e-business applications, by size class, EU, 2021
(% of enterprises)

	Variables	All enterprises	Small	Medium	Large
Enterprises with a website	e_web	78	75	89	94
Use of Enterprise Resource Planning (ERP)	e_erp1	38	33	62	81
Use of Customer Relationship Management (CRM)	e_crm	35	31	50	65

Notes: 2021 EU aggregates for website variables excluding Belgium, Denmark and Estonia.

Source: Eurostat (isoc_ciweb) and (isoc_eb_iip)

6 Sectoral analysis of the electronic business of companies in the countries of the European Union

In any case, the development of electronic business of companies in the countries of the

European Union is different in individual sectors, partly caused by differences in the very nature of business. This is clearly seen from the data presented in Table 13.

Table 13. Enterprises having ERP software package, and Enterprises using Customer Relationship Management (CRM), by economic activity, EU, 2021 (% enterprises)

	Enterprises who have an ERP software package	Enterprises using analytical CRM	Enterprises using operational CRM
	e_erp1	E_CRMAN	E_CRMSTR
Information and communication	54	42	63
Manufacturing	49	36	44
Electricity, gas, steam and air conditioning; water supply, sewerage, waste management and remediation activities	48	17	43
Wholesale and retail trade; repair of motor vehicles and motorcycles	46	19	42
Real estate activities	44	28	41
Professional, scientific and technical activities	39	16	40
Retail trade	33	18	35
Administrative and support service activities	31	19	34
Transport and storage	30	21	29
Accommodation	28	12	24
Construction	26	7	22
All activities	38	19	34

Source: Eurostat (isoc_eb_iip)

The highest percentage of Enterprises who have ERP software package, Enterprises using analytical CRM and Enterprises using operational CRM is in the Information and communication sector. It is the smallest in the Construction sector. In the Wholesale and retail trade sector; repair of motor vehicles and motorcycles and Retail trade electronic business is at an enviable level. The character itself, among other things, affects the level of development of electronic business of companies from different sectors. Investment needs in information and communication technology are different for each sector.

7 Conclusion

Empirical research on electronic business of companies in the European Union and Serbia leads to the following conclusions:

In the countries of the European Union, the trend is to increase the electronic business of companies. Thus, for example, in 2021 compared to 2019, there was an increase in Enterprises with a website, Use of Enterprise Resource Planning (ERP) and Use of Customer Relationship Management (CRM). According to the results of the LMAW-DNMA method, the top five companies of the European Union countries in terms of the development of electronic business are in order: Spain, Netherlands, Austria, France and Cyprus. As far as the leading countries of the European Union are concerned, the positioning of companies from Germany (eighth place), France (fourth place) and Italy (sixteenth place) is at an enviable level. The worst position was taken by Estonian companies. Romanian companies are positioned on the twenty-third place.

As far as Serbian companies are concerned, they are positioned in twenty-fifth place. They are positioned worse in relation to the companies of the European Union, and in the region, namely Croatia (twenty-second place) and Slovenia (seventeenth place). In Serbia, the percentage of companies that use the website is higher than in North Macedonia and Bosnia and Herzegovina. However, compared to Croatia and Slovenia, the percentage of companies in Serbia that use the website is lower. The percentage of companies in Serbia that use *Use of Enterprise Resource Planning (ERP)* is lower than in Croatia and Slovenia. Compared to North Macedonia, it is larger, but smaller compared to Bosnia and Herzegovina. In Serbia, the percentage of companies that use Use of Customer Relationship Management (CRM) is lower than in Croatia and Slovenia. It is also smaller compared to North Macedonia and Bosnia and Herzegovina. Unsatisfactory level of development of electronic business of companies in Serbia, especially in relation to developed countries of the European Union. In the future, due to its importance, it is necessary for companies in Serbia to develop and use electronic business as much as possible in their operations, and for these needs to invest in modern information and communication technology.

In the European Union, the development of electronic business of companies varies by size. It is significantly higher in large companies than in small and medium-sized ones. Likewise, it is higher in medium-sized than in small enterprises. The size of the company therefore significantly affects the level of application of electronic business. Compared to small and larger companies, they are also in a better position to invest in information and communication technology. The highest percentage of Enterprises who have ERP software package, Enterprises using analytical CRM and Enterprises using operational CRM is in the Information and communication sector, and the lowest is in the Construction sector. In the Wholesale and retail trade sector; repair of motor vehicles

and motorcycles and Retail trade electronic business is at an enviable level. Therefore, the sam sector's character, among other things, affects the level of development of the company's electronic business, as well as the needs for investment in information and communication technology.

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