OBJECTIVE ANALYSIS AND RANKING OF HUNGARIAN CITIES, WITH DIFFERENT CLASSIFICATION TECHNIQUES, PART 2: ANALYSIS

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Összefoglalás – A tanulmány célja, hogy a magyarországi városokat és megyéket környezetminőségük és környezeti tudatosságuk szintje alapján osztályozza. Bemutatjuk a magyarországi városok és megyek rangsorát azok „Green Cities Index”, illetve a „Green Counties Index” értékeinek összevetésével. Az 1. részben (*Makra and Sümeghy*, 2007) bemutatott módszertan szerint a városokat, illetve megyéket eltérő klasszifikációs technikák szerint osztályoztuk, s elemeztük az osztályozás hatékonyságát. Azonban ezek egyike sem adott elfogadható eredményt sem a városokra, sem a megyékre. E három algoritmus paraméterei alapján egyik clusterezési eljárás során sem találtunk elfogadható cluster-szerkezetet. A *fanny* algoritmus alkalmazásával kapott clusterek – jóllehet gyenge a szerkezetük – kiterjedt és jól körülhatárolható térségeket jeleznek Magyarországon, melyek adott földrajzi objektumokkal jól körülírhatók.

Summary – The aim of the study was to rank and classify Hungarian cities and counties according to their environmental quality and level of environmental awareness. The rankings of the Hungarian cities and counties are based on their „Green Cities Index” and „Green Counties Index” values. According to the methodology presented in Part 1 (*Makra and Sümeghy*, 2007), cities and counties were grouped with different classification techniques and the efficacy of the classification was analysed. However, these did not give acceptable results for the cities, nor for the counties. According to the parameters of the here-mentioned three algorithms, no reasonable structures were found in any clustering. Clusters received applying the algorithm *fanny*, though having weak structure, indicate large and definite regions in Hungary, which can well be circumscribed by geographical objects.

Key words: environmental indicators, Green Cities Index, Green Counties Index, ranking, factor analysis, clustering, SPSS-software, R-language; algorithms: agnes, fanny, pam

1. RESULTS

1.1. Ranking

1.1.1. Cities

The final sequence of the cities shows some surprising results (*Table 2*). Nagykanizsa, near the Hungarian-Croatian border, is the highest-ranked city. It is followed by settlements around Lake Balaton: Balatonföldvár (2), Balatonboglár (3) and Balatonlelle (4). Among the major cities, Szombathely (5), Zalaegerszeg (7) and Kaposvár (8) stand out (*Table 2*).

Mosonmagyaróvár (88), Mór (87) and Balassagyarmat (86) are the worst ranked cities (*Table 2*) in spite of their relatively good rank in a number of indicators. Summing up, no city is found consistently either at the top or the bottom half of the rankings on all

environmental indicators. All cities in Hungary are characterised by a mix of favourable and less favourable environmental quality.

The environmental quality of Hungarian cities is best in the western and southern parts of Transdanubia, where Green Cities Index values are smallest. There are no clear regional patterns in the rest of the country (*Fig. 1*).

Fig. 1 Environmental quality of cities according to their Green Cities Index
[High values (circles with large area) = favourable; Low values (circles with small area) = disadvantageous]. The numbers indicate the final sequence of the cities
(1 = best, 88 = worst).

Fig. 2 Environmental quality of counties according to their Green Counties Index
[High values (circles with large area) = favourable; Low values (circles with small area) = disadvantageous]. The numbers indicate the final sequence of the counties
(1 = best, 19 = worst).

1.1.1.1. Potential impact of population on the Green Cities Index

The possible consequence of including population and population density in the Green Index was examined by comparing the rankings obtained with the inclusion of the two variables (modified Final Sequence) and those calculated without them (original Final Sequence). The Spearman's rank correlation coefficient, which was utilized for this purpose, yielded a value of 0.94 significant at the 99.9% confidence level. This means that there is a significant connection between the original and modified groups of indicators. We would be in error once in 1000 cases. Hence, the original final sequence is not substantially influenced by not considering population and population density. This result indicates that, although not perfect, the Green Cities Index, as calculated, is a reasonably fair method of providing an environmental rating for cities in Hungary.

1.1.2. Counties

According to the final rank order of the counties (*Table 1*), Somogy is the greenest county of Hungary. Though it is almost the most wasteful in water consumption (ranked 18) and average in waste removal (13), its favourable ranking in public green area total (1), average sulphur dioxide concentration (1), energy requirement and electric energy consumption (2 and 4, respectively), regularly cleaned constructed public surfaces (3) and average concentration of particulates deposited (3) make it the most environment-friendly county in the country. Somogy is followed by Zala and Vas respectively. Both Zala and Vas score well in environmental factors related to infrastructural and social developments and to a lesser extent, in physical factors such as air quality and green areas. The Green

Counties Index is a good measure of the general level of development of the counties. It well reflects the fact that the western part of the country, namely Transdanubia, is much more environment-sensitively developed than eastern Hungary.

The seven best counties are all found in Transdanubia (Somogy, Zala, Vas, Komárom-Esztergom, Veszprém, Baranya and Győr-Moson-Sopron), while the worst five all in the Great Hungarian Plain: Szabolcs-Szatmár-Bereg, Hajdú-Bihar, Békés, Bács-Kiskun and Heves. However Szabolcs-Szatmár-Bereg, Hajdú-Bihar and Békés do well in some indicators.

Table 1 Average of rankings of the environmental indicators considered, namely the Green Counties Index, and the final sequence of the counties (1 = best, 19 = worst; the numbers indicate the counties)

County	Final sequence	Green Counties Index
Somogy	1	6.32
Zala	2	6.79
Vas	3	7.21
Komárom-Esztergom	4	8.79
Veszprém	5	9.26
Baranya	6	9.53
Győr-Moson-Sopron	7	10.00
Pest	8	10.05
Fejér	9	10.16
Jász-Nagykun-Szolnok	10	10.16
Borsod-Abaúj-Zemplén	11	10.42
Csongrád	12	10.47
Tolna	13	10.79
Nógrád	14	11.00
Heves	15	11.21
Bács-Kiskun	16	11.47
Békés	17	11.58
Hajdú-Bihar	18	11.89
Szabolcs-Szatmár-Bereg	19	12.32

Like in the case of cities, there is no county found consistently at either the top or the bottom half of the rankings, in all indicators. In general, Transdanubian counties have better ranks than counties in eastern Hungary (*Fig. 2*).

1.2. Clustering procedures

1.2.1. Cluster analysis using SPSS-software

After performing factor analysis, cluster analysis is applied to the factor scores time series in order to objectively group cities and counties according to their similar characteristics. In this paper, the agglomerative hierarchical technique, *Ward's* method is applied.

1.2.1.1. Cities

After performing cluster analysis, the 88 cities were divided into 6 groups, considered to be the most homogenous. The groups received do not form a comprehensive (contiguous) spatial system (*Fig. 3*). All the 14 cities of Group 1 are found either in eastern or northern Hungary, indicating considerable dispersion. Group 2 consists of 6 Transdanubian settlements, 4 are located in the southwestern part of Transdanubia, while the other two are far from them. The 30 cities of Group 3 also exhibit considerable spatial dispersion. Here, two distinct sub-groups are found; one in the southern part of Transdanubia and the other in the northern part. Four cities in Group 4 are found around Lake Balaton, while the other two are in the southern part of the Great Hungarian Plain. All the 4 cities of Group 5 are found around Budapest. Though settlements belonging to Group 6 (28 cities) show density junctions in the middle part of Transdanubia, south of Budapest and Northern Hungary they are considerably dispersed (*Fig. 3*).

Table 2 Average of rankings of the environmental indicators considered namely, the Green Cities Index, and the final sequence of the cities (1 = best, 88 = worst) City

City	Final sequence	Green Cities Index	City	Final sequence	Green Cities Index	City	Final sequence	Green Cities Index
Nagykanizsa	1	29.89	Siklós	31	42.26	Esztergom	61	47.74
Balatonföldvár	2	30.58	Szeged	32	42.32	Göd	62	47.78
Balatonboglár	3	30.68	Vác	33	42.47	Dombóvár	63	48.16
Balatonlelle	4	32.11	Dorog	34	42.74	Békés	64	48.37
Szombathely	5	32.89	Debrecen	35	42.84	Lőrinci	65	48.84
Tiszaújváros	6	33.00	Szigetvár	36	42.88	Nagymaros	66	50.00
Zalaegerszeg	7	33.16	Bátonyterenye	37	43.00	Cegléd	67	50.74
Kaposvár	8	33.32	Baja	38	43.26	Hajdúnánás	68	50.84
Siófok	9	34.32	Tata	39	43.32	Pápa	69	51.05
Százhalombatta	10	34.63	Mohács	40	43.95	Szentendre	70	51.14
Fonyód	11	34.74	Gyöngyös	41	44.05	Sümeg	71	51.32
Bonyhád	12	34.79	Békéscsaba	42	44.21	Szécsény	72	51.42
Tapolca	13	35.95	Kőszeg	43	44.37	Komárom	73	51.47
Tatabánya	14	36.26	Ózd	44	44.53	Ajka	74	51.58
Miskolc	15	36.84	Gyula	45	44.63	Pásztó	75	51.63
Komló	16	37.16	Balatonfüred	46	44.79	Mátészalka	76	51.74
Oroszlány	17	37.21	Salgótarján	47	44.89	Budaörs	77	52.00
Lenti	18	37.68	Jászberény	48	45.00	Záhony	78	52.11
Szolnok	19	38.26	Hajdúszoboszló	49	45.05	Szentlőrinc	79	52.37
Győr	20	38.58	Dunakeszi	50	45.47	Orosháza	80	53.16
Sopron	21	39.05	Hatvan	51	45.63	Kisvárda	81	53.26
Kazincbarcika	22	39.37	Veresegyház	52	46.00	Zirc	82	54.16
Budapest	23	39.74	Balatonalmádi	53	46.21	Kistelek	83	54.26
Székesfehérvár	24	39.89	Kalocsa	54	46.27	Tiszavasvári	84	55.89
Szekszárd	25	39.91	Várpalota	55	46.58	Sajószentpéter	85	56.79
Pécs	26	40.00	Kecskemét	56	46.74	Balassagyarmat	86	58.05
Keszthely	27	40.32	Nyíregyháza	57	46.95	Mór	87	58.68
Eger	28	40.74	Gárdony	58	47.21	Mosonmagyaróvár	88	59.21
Dunaújváros	29	41.21	Csongrád	59	47.26			
Pilisvörösvár	30	41.63	Veszprém	60	47.42			

1.2.1.2. Counties

The grouping of counties using cluster analysis separated regions more clearly. The southern part of the Great Hungarian Plain (Bács-Kiskun, Csongrád and Békés counties) is well defined. The middle part of the Great Hungarian Plain and Northern Hungary (Jász-Nagykun-Szolnok, Hajdú-Bihar, Nógrád, Heves, Borsod-Abaúj-Zemplén and Szabolcs-

Szatmár-Bereg counties) stand out as well. Zala, Somogy and Fejér counties form a distinct region as do regions representing Vas, Veszprém and Komárom-Esztergom counties; Baranya and Tolna; Győr-Moson-Sopron and Pest (*Fig. 4*).

Fig. 3 Spatial distribution of cities, with symbols of their 6 clusters, using cluster analysis, agglomerative method, after performing factor analysis, SPSS.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.]

Fig. 4 Spatial distribution of counties, with symbols of their 6 clusters, using cluster analysis, agglomerative method, after performing factor analysis, SPSS.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.]

1.2.2. Cluster analysis using R-language

When applying the agglomerative method, the goodness of clustering is indicated by the agglomerative coefficient (*AC*). The higher the *AC*, the better the clustering is. However, it should be emphasized again, that the *AC* tends to increase with the number of objects.

The application of classification techniques did not result in strong structure either for the cities or the counties. The silhouette coefficient (*SC*) of methods *fanny* and *pam* shows values between 0.00 and 0.50. Among them the only highest ones were retained. In spite of this, they refer to the weak structure of the database (*SC* is between 0.26 and 0.50) (see *Table 2* in *Chapter 3.3.1.1.* in *Makra and Sümeghy* (2007)).

1.2.2.1. Cities

Firstly the algorithm *agnes* was applied to the database comprising 19 variables for each city. As a result of this analysis 7 clusters were received (*Fig. 5*). Spatial distribution of the cities with symbols of their 7 clusters does not indicate any clearly homogenous regions. However, definite subregions of cities belonging to cluster 1 can be observed in South-Transdanubia and, along a SW-NE axis, in North-Transdanubia. Furthermore, cities of cluster 5 are predominant in Western Transdanubia (*Fig. 5*).

Afterwards, factor analysis was used on the original database (to the 19 environmental indicators of the 88 cities) and then the algorithm *agnes* was applied to the seven factor score time series received. Altogether 7 clusters were established (*Fig. 6*). The spatial distribution of the cities with symbols of their 7 clusters also shows homogenous subregions. They are as follows: South Transdanubia with cities of cluster 7, SW Transdanubia with cities belonging to cluster 6 and Lake Balaton region with cities of

cluster 5. Smaller homogenous regions are North Transdanubia with cities of cluster 7, North Hungary with cities of cluster 1 and the southern part of the Great Hungarian Plain with cities of cluster 5 (*Fig. 6*).

Fig. 5 Spatial distribution of cities, with symbols of their 7 clusters using agnes algorithm based on the 19 environmental indicators.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.] (AC = 0.80)

Fig. 6 Spatial distribution of cities, with symbols of their 7 clusters, using agnes algorithm, after performing factor analysis, SPSS.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.] (AC = 0.84)

Algorithms *agnes*, *fanny* and *pam* did not give acceptable results ($SC \leq 0.25$) neither for the database comprising 19 variables of the 88 cities, nor for their application to the seven factor score time series after performing factor analysis.

1.2.2.2. Counties

Firstly, the algorithm *agnes* was applied to the database including 19 variables for each county. This analysis resulted in 8 clusters (*Fig. 7*). Spatial distribution of the counties with symbols of their 8 clusters indicates definite regions with similar characteristics. They are counties of cluster 2, which cover all Eastern and Northern Hungary, furthermore counties of cluster 5 which are found in Southern and NNW Transdanubia. However, it should be mentioned that clusters 1, 4, 7 and 8 include only one county each (*Fig. 7*).

Then, factor analysis was used on the original database (to the 19 environmental indicators of the 19 counties) and after that the algorithm *agnes* was applied to the seven factor score time series received. Altogether 8 clusters were established (*Fig. 8*). Spatial distribution of the counties with symbols of their 8 clusters shows different homogenous regions. The largest of them with four counties of cluster 5 is found in the middle part of the Great Hungarian Plain and in NE Hungary. Though cluster 3 comprises altogether 6 counties, it is divided into three two-county subregions. Moreover, clusters 1, 5, 7 and 8 include only one county each (*Fig. 8*).

Algorithms *agnes*, *fanny* and *pam*, similarly to the case of the cities, did not give acceptable results ($SC \leq 0.25$) neither for the database including 19 variables of the 19 counties, nor for their application to the seven factor score time series after performing factor analysis.

Thereafter, algorithm *mona* was applied to the database including the 19 variables. The banner belonging to this analysis is found on *Fig. 9*. It indicates 15 clusters; hence, this result is omitted from further consideration. Furthermore, the banner shows that the

algorithm classified the whole database using 8 variables. This result gave an idea to use the algorithms *agnes*, *fanny* and *pam* on these 8 variables. Then, factor analysis was applied to these 8 variables. According to the *Guttmann criterion* 3 factors were retained, for which algorithms *agnes*, *fanny* and *pam* were performed again.

Fig. 7 Spatial distribution of counties, with symbols of their 8 clusters, using agnes algorithm based on the 19 environmental indicators.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.] (AC = 0.61)

Fig. 8 Spatial distribution of counties, with symbols of their 8 clusters, using agnes algorithm, after performing factor analysis, SPSS.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.] (AC = 0.58)
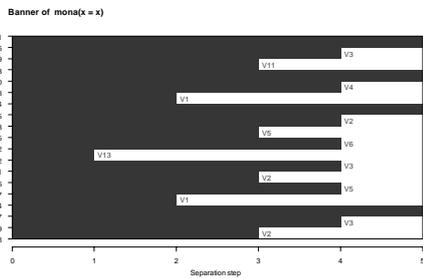

Fig. 9 Banner of function *mona* based on the 19 environmental indicators, counties

Fig. 10 Spatial distribution of counties, with symbols of their 7 clusters, using agnes algorithm, after performing mona algorithm.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.] (AC = 0.65)

The analysis using algorithm *agnes*, after performing algorithm *mona* resulted in 7 clusters (*Fig. 10*). The largest region belongs to cluster 1 comprising three uniform counties in Southern Hungary, while the fourth county belonging to cluster 1 is found in North Hungary. Clusters 3 and 4 are divided into 4 and 2 subregions, respectively (*Fig. 10*). Thereafter, algorithm *agnes* was used after performing algorithm *mona* and then factor analysis was performed to the three factor score time series retained. As a result of this analysis, 5 clusters were received (*Fig. 11*). The spatial distribution of the counties with

symbols of their 5 clusters indicates a large uniform region in Eastern Hungary with five counties of cluster 3. Another extended region is characterized by cluster 5; however, it is divided into two parts: a three-county subregion in North Transdanubia and a two-county subregion in North Hungary (*Fig. 11*).

Algorithm *fanny*, which was applied to the database comprising 8 variables, did not give acceptable clustering. The analysis applied to the three retained factors resulted in only one clustering, namely for k = 5, for which *SC* > 0.25 (*Fig. 12*). As a result of the analysis, 4 clusters were received only. Spatial distribution of the counties with symbols of their 4 clusters indicates large and definite regions in Hungary. Eastern Hungary is characterized by five counties of cluster 3, South Transdanubia by four counties of cluster 2, North Transdanubia by three counties of cluster 4 and Northern Hungary by three counties of cluster 1 (*Fig. 12*).

Fig. 11 Spatial distribution of counties, with symbols of their 5 clusters, using agnes algorithm after performing mona algorithm and then factor analysis.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.] (AC = 0.83)

Fig. 12 Spatial distribution of counties, with symbols of their 4 clusters, using fanny algorithm (with k = 5) after performing mona algorithm and then factor analysis.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.]
[Average silhouette width (*ASW*) = 0.32]

Applying algorithm *pam* to the database comprising 8 variables did not give acceptable clustering (*SC* ≤ 0.25). Algorithm *pam* applied to the three factor score time series after performing factor analysis to the 8 variables, gave acceptable results only for the cases k = 5 (*Fig. 13*), k = 7 (*Fig. 14*) and k = 8 (*Fig. 15*) (*SC* > 0.25). The spatial distribution of counties with symbols of their 5 clusters shows two large uniform regions. The largest one is found in Eastern Hungary with five counties of cluster 3. However, one county in the middle part of Transdanubia also belongs to this cluster.

The other important region is that of cluster 4, which is divided into two parts: North Transdanubia with four counties and Northern Hungary with two counties (*Fig. 13*). The uniform regions indicated by counties with symbols of their 7 clusters become smaller. Regions in Middle and East Hungary with three counties of cluster 3 and North Transdanubia with four counties of cluster 6 are the most characteristic (*Fig. 14*). Spatial distribution of counties with symbols of their 8 clusters shows smaller uniform regions than that for k = 7. The largest uniform regions are found in Middle and East Hungary with three

counties of cluster 3 and in North Transdanubia with also three counties of cluster 6 (*Fig. 15*).

Fig. 13 Spatial distribution of counties, with symbols of their 5 clusters, using pam algorithm (with k = 5) after performing mona algorithm and then factor analysis.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.]
[Average silhouette width (*ASW*) = 0.33]

Fig. 14 Spatial distribution of counties, with symbols of their 7 clusters, using pam algorithm (with k = 7) after performing mona algorithm and then factor analysis.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.]
[Average silhouette width (*ASW*) = 0.37]

Fig. 15 Spatial distribution of counties, with symbols of their 8 clusters, using pam algorithm (with k = 8) after performing mona algorithm and then factor analysis.
[Right and down the sign, serial number of the cluster and the number of cities in the cluster (in parentheses) are found.]
[Average silhouette width (*ASW*) = 0.35]

2. CONCLUSION

The aim of the study was to rank and classify Hungarian cities and counties according to their environmental quality and level of environmental awareness.

The top 5 most environmentally friendly cities are, in descending order, Nagykanizsa, Balatonföldvár, Balatonboglár, Balatonlelle and Szombathely. The bottom

five are, starting with the worst, Mosonmagyaróvár, Mór, Balassagyarmat, Sajószentpéter and Tiszavasvári. Cities situated in the western and southwestern part of Transdanubia have the best environmental quality. In the rest of the country, cities with either favourable or unfavourable positions are mixed, forming no comprehensive regional patterns.

The top 3 counties are Somogy, Vas and Zala; Szabolcs-Szatmár-Bereg, Hajdú-Bihar and Békés counties are the most disadvantaged. The most environment-friendly counties can be found in Transdanubia, clearly separated from the least environment-friendly ones found in eastern Hungary.

Clustering was performed with cluster analysis using both SPSS software and R-language. Cluster analysis with the application of SPSS software resulted in 6 most homogenous groups of cities, which did not form comprehensive spatial patterns. The classification of the counties according to cluster analysis determined also 6 clear groups of them.

Cluster analysis using R-language was carried out with different procedures. Algorithms *agnes*, *fanny* and *pam* did not give acceptable results ($SC \leq 0.25$) neither for the database of the cities, nor for the counties. The silhouette coefficient did not exceed the value 0.5 in either case, which means that a reasonable structure was found. Clusters received applying algorithm *fanny*, though having weak structure, indicate large and definite regions in Hungary, which can be circumscribed by clear geographical objects. The *agglomerative coefficient* (*AC*), which measures the goodness of the clustering of the dataset, shows the highest values when (1) clustering cities with 7 clusters, using algorithm *agnes* (*AC* = 0.80; *Fig. 5*), (2) clustering cities with 7 clusters, using algorithm *agnes* after performing factor analysis, SPSS (*AC* = 0.84; *Fig. 6*) and (3) clustering counties with 5 clusters, using algorithm *agnes* after performing algorithm *mona* and then factor analysis (*AC* = 0.83; *Fig. 11*).

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