Governance networks around grasslands with contrasting management history

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Abstract

Romanian grasslands have high nature value, being between the best biodiversity hotspots at the

global level. Current agri-environment measures of the European Union Common Agricultural

Policy (CAP) contradicts the Biodiversity Strategy to 2020 objective by hindering coordinated

grassland governance and collaboration among the involved actors. At the European level, few

attempts have been made in creating conceptual strategies for implementing conservation

measures in a multi-actor and multi-scale governance setting. Our paper focuses on a

comparative network analysis of grassland landscape governance of three Romanian regions

(Iron Gates Natural Park – SW; Sighisoara - Tarnava Mare – center; and Dobrogea - SE),

representatives for grassland management in mountain and lowland settings. In Sighisoara,

grasslands governance has been centralized but biodiversity-friendly, while in Iron Gates,

grasslands were traditionally managed through a decentralized, community-level system, and this

type of governance continues to date. Whereas for Dobrogea's grasslands, the governance was

performed in an intensive, centralized state-run management regime during the communist time

and by large landowners after the transition period ended. Our findings illustrate the structure of

the three governance networks and dissimilar patterns of collaboration, indicating distinct

particularities to be considered when exploring barriers to and options for successful governance

in traditionally managed grasslands in the context of CAP measures-driven management.

Keywords: Common Agricultural Policy, network analysis, governance, Iron Gates, Macin

Mountains, Sighisoara

Introduction

Semi-natural grasslands of Europe have high economic, socio-cultural, and natural values (Plieninger et al., 2015). Such landscapes influenced local economies and communities (Bergmeier et al., 2010) and favored the emergence of informal institutions that shaped particular norms, rules, knowledge, and skills (O'Farrell and Anderson, 2010). One key driver of the erosion of multifunctional grassland management in Europe was the establishment of the formal institutions governing the management of these landscapes. Notably, with the economic development of the agricultural and forestry sectors, a shift from the traditional multifunctional management towards monofunctional intensive management of grasslands happened as a response to the increasing social demand for agricultural and timber products (Hartel et al., 2015). As a result of this shift, the importance of the traditional (informal) institutions in the multifunctional management of the grasslands was weakened by the formal (i.e., state-based) institutions (Hartel et al., 2016). The prevalence of formal institutional governance coupled with the monofunctional grassland management resulted in a sharp decline of hay meadows and pastures with high socio-cultural and natural values across Europe, especially in the economically developed countries (Overmars et al., 2013; Pe'er et al., 2014). Despite the declining trends in the European continent, traditionally managed grasslands are still present in countries where traditional management practices persist mainly as a legacy of the geopolitical instability (Plieninger et al., 2015). Europe's last traditionally managed grasslands provide a unique opportunity to understand the institutional network involved in the governance of these systems mainly because contradictions exist between the 2014-2020 Common Agricultural Policy (CAP) measures and the European Union (EU) biodiversity Strategy to 2020 (Balázsi, 2018; Beaufoy et al., 2015). These contradictions also create conflicts between the users of these

grasslands (Milczarek-Andrzejewska et al., 2018). This situation can be interpreted as a case of

"rigidity trap" (Carpenter and Brock, 2008), where the landowners recognize the unsuitability of

CAP measures for the grassland social-ecological system (Leventon et al., 2017) but are

nevertheless encouraged to implement even undesired practices in order to receive EU subsidies.

Furthermore, the acquisition of agricultural lands by large landowners (land concentration)

allows a small number of owners to control large swaths of land (Calota and Patru-Stupariu,

2019; Constantin et al., 2017). Such threats may simplify the management of grasslands, change

traditional connections, and disrupt traditional land use (Balázsi, 2018; Hartel and Plieninger,

2014).

The governance structures involved in landscape management are often fragmented because

policies and operational responsibilities are divided between an array of competing organizations

and individuals, often with contradictory objectives (Gieseke, 2020). This is also the case of

Romania, for example, in protected areas and water resource governance (Manolache et al.,

2018; Nita et al., 2018; Vinke-de Kruijf et al., 2016). The complex governance arenas make the

analysis of stakeholders' interaction and response to policies difficult with conventional tools,

and network analysis may provide an ideal framework to understand the formation and

coordination of management structures and provide feedback to policymakers (Alexander et al.,

2016; Bodin et al., 2019; Nita et al., 2018).

Considering the complex socio-cultural and institutional background around multifunctional

management, our goal is to employ social network methods to understand governance systems

around the management of ancient grasslands from Romania and provide insights for the

multilevel governance strategies of these systems. Network governance has become an

increasingly used approach among practitioners from various research areas: collaborative public

management, resource governance, ecology, conservation planning, environmental policy

(Berardo et al., 2014; Bodin et al., 2016; Manolache et al., 2018). The investigation of

governance networks allows the analysis of informal and formal arrangements where people or

organizations work together towards a common goal (Scarlett and McKinney, 2016). We refer to

people or organizations active in the governance networks as institutions performing repetitive

and structured interactions (Ostrom, 2005) with the aim of managing grasslands at all spatial

scales.

We focus on a comparative investigation of network structures around the multifunctional

grasslands from three traditional regions of Romania, all three covered by protected area

regulations. Based on the differences related to the history and socio-cultural context of the three

regions (see below), we hypothesize that there will be significant differences between the

governance networks from the three investigated areas, especially related to leadership structure

under CAP influence. We tested the following research hypotheses: 1) The grassland governance

networks developed under CAP influence are influenced by the current administrative context

and not the past farming practices, expecting from this point of view more centralized networks

in regions with fewer administrative and territorial jurisdictions. 2) Direct payments under the

EU CAP impedes the emergence of local stakeholder networks (informal institutions) around

grassland governance. 3) The past informal institutions (such as the communal and family

farming types) have a strong influence on the ways how farmers sense the grassland ownership

in the present and on the ways how they organize to better capitalize on the EU CAP payments.

Methods

Case studies

We selected three biocultural regions as case studies capturing different environments and past farming practices, as follows: a) the region of Macin Mountains (Macin Mountains National Park, hereafter Macin Mountains) within Dobrogea – South-East Romania, illustrative for grassland management in lowland areas. b) The Saxon cultural region from Central Romania (the area covered by Sighisoara-Tarnava Mare Natura 2000 site, hereafter Sighisoara area, the town of Sighisoara is situated in the central area of the region) which is illustrative for hilly (tableland) areas and c) The cultural region of Banat (the area covered by Iron Gates Natural Park, hereafter Iron Gates) from South-Western Romania, illustrative for low elevation mountains areas (Supplementary Figure 1). A centralized, state-run management regime characterized Macin Mountains (Dobrogea) during the communist time. After the transition period ended, Macin Mountains grasslands were managed primarily by large landowners with large collective farms, dairy farms, with little ownership sense (Anthony and Moldovan, 2008; MMNPA, 2013). Sighisoara area is known as a tableland with diverse land-use and a mixture of communal and private small farms (Fischer et al., 2012; Sutcliffe et al., 2013). In contrast, Iron Gates from South-West Romania is situated at the border with former Yugoslavia, characterized by low altitude mountains, small private farms with diverse land-use (e.g., small patches of corn plantation, hayfields, small orchards). In Iron Gates, grassland management was traditionally performed in a decentralized, community-level system, and this type of management continues to this day (Pătroescu and Rozylowicz, 2000). The three protected areas overlap one county and 18 local administrative units in Macin Mountains, two counties and 15 local administrative units in Iron Gates, and three counties and 14 local administrative units in the Sighisoara area (see Supplementary Figure 1).

Network data

We collected network data by face-to-face semi-structured interviews applied in 2018 with

representatives of institutions potentially involved in the management of grassland in the three

protected areas. To select the interviewed institutions, we reviewed Romanian legal norms and

CAP technical documents related to grasslands and extracted local, regional, and national likely

to be involved in the governance of grassland in the three protected areas. This step resulted in a

predefined list of institutions for each case study. If the interviewed persons nominated actors not

included in the predefined list, we applied the interviews to key representatives of these

institutions. We targeted institutions involved in contribution to information flow in grassland

management (e.g., informing about direct payments requirements, grassland management

technologies, grassland biodiversity, and conservation, funding availability, agri-environment

schemes), control of grassland management (e.g., complying with requirements of direct

payments, complying with various legal norms), approval of direct payments (submission of

direct payments documentation, payments to farmers). The list of interviewed institutions

included local, regional, and national agencies and authorities, local and regional

administrations, private and public enterprises, NGOs, local associative organizations, resource

managers, active in grassland management. We did not interview individual farmers and national

agencies. The farmers were included in the resulted network as a collective institution. National

agencies/authorities were not interviewed because they have only a top-down/indirect influence

on the network, their competencies being delegated to local representatives already included in

the network (e.g., Ministry of Agriculture to the local Agencies for Payments and Intervention in

Agriculture).

The respondents were asked to indicate other institutions with whom they were collaborating, the

direction of collaboration (e.g., they send information to the named institution, they receive

information from the named institution or both), type of exchanged information, and the

frequency of contact. We considered as exchanged information any regulatory/administrative

task, joint projects, participation in the same working groups, drafting reports and studies in

common, other shared activities. The semi-structured interviews included a question about the

quality of collaboration on a Likert 5 points scale. However, the representatives of institutions

were reluctant to provide an answer; thus, we discarded the data from analysis.

For each protected area, we constructed a grassland governance network, where institutions

(vertices) were linked (edges) if they transfer any information related to grassland management.

Because we included in the analysis the direction of information transfer (sender institution,

receiver institution), the networks are one-mode directed networks (Barabási, 2016; Borgatti et

al., 2017). Furthermore, we classified the institutions as (1) public authorities (ministry, local,

regional, and national governmental agencies), (2) public administration (city halls), (3) NGOs,

and research and education bodies, and (4) enterprises (i.e., industry, private businesses, state-

owned businesses, farmers).

Network-level analyses

To investigate if the structural characteristics of grassland governance networks in protected

areas are influenced by farming management history and local administrative complexity, we

used standard cohesion and reciprocity metrics and exponential random graph models for one-

mode directed network (Barabási, 2016; Borgatti et al., 2017; Hanneman and Riddle, 2005;

Lusher et al., 2013).

The *cohesion and reciprocity metrics* for directed networks considered here were: size, volume,

average degree, network density, network fragmentation, network diameter, average shortest

path, network compactness, and arc reciprocity (Borgatti et al., 2017; Hanneman and Riddle,

2005). These metrics allow us to infer which case study displays a higher level of interaction

between the actors and an efficient information flow. The *size of the network* is represented by

the number of institutions participating in the respective network, while the *network volume* by

the number of edges or connections between institutions (Barabási, 2016). In our case, both

metrics are influenced by the administrative complexity (e.g., number of counties overlapping

the analyzed protected area), protected areas overlapping more countries having larger and more

complex governance networks. *Network density* represents the proportion of observed

connections between institutions participating in transferring information to the maximum

number of possible connections in the respective network. This metric indicates the magnitude of

interconnectivity between network members (networks with densities closer to 1 are dense, while

those with densities closer to 0 are sparse) (Borgatti et al., 2017). *Arc reciprocity* represents the

proportion of mutual connections between organizations divided by the total number of

connections (reciprocal or otherwise). A network high arc reciprocity indicates a high proportion

of bidirectional connections between network actors (Borgatti et al., 2017). The degree of a node

(institution) is the number of connections of the respective node regardless of the direction of

information transfer. The *average degree* of the network can be used to compare networks, e.g.,

networks with low density and relatively high average degree indicate a centralization around a

small number of highly connected network members (Barabási, 2016). Average path length and

network diameter can be used to assess the information flow in the network. *Average path length*

represents the average number of institutions that must be passed to connect two institutions in

the network, and the *network diameter* the shortest path length between the two most distant

nodes in the network. Information flows faster and more effectively in networks with small

average path length, while in networks with small diameter, distant members are well connected

(Barabási, 2016). A closely related metric is the *fragmentation* of the network, which measures

the proportion of pairs of institutions that cannot reach each other by a path of any length. Higher

fragmentation indicates a limited potential for exchanging information (Chen et al., 2007).

We use exponential random graph models (ERGM or p* models) to infer if observed network

substructures (configurations) are more commonly observed in the network than expected by

chance (Lusher et al., 2013; Robins et al., 2012; Wang et al., 2009). ERGMs are related to

general linear models, modified to deal with the lack of independence of observations. More

details ERGM statistics and assumptions are provided by Wang et al. (2009). ERGM for directed

networks can accommodate several configurations such as density, reciprocity, k-in-stars, k-out-

stars, 2-mixed-stars, alternating-in-star (Wang et al., 2009). To fit the ERGM for each

governance network, we use P-Net for directed network, as suggested by Lusher et al. (2013): 1)

we selected several configurations to be estimated, run the model with default values, and

checked the model statistics. Density for all estimations was fixed (i.e., the observed network

density); 2) if *t*-ratios for selected network configurations were less than 4 for all values, we

updated the parameters with the new estimated values, and re-run the model using an increased

multiplication-factor. Because the analysis is trial and error, we repeated the procedure until the

model converged or we included/excluded configurations and repeated the first 2 steps if the

model was unreliable; 3) if the model converged (all selected parameters had *t*-statistics <0.1 in

absolute values) we estimated the Goodness-of-Fit using 100,000,000 simulations. For

Goodness-of-Fit analysis, we updated the converged parameters with the values obtained in step

2 and zero for estimates for all other parameters. A model with a good fit has the absolute value

of the *t*-ratios <0.1 for converged parameters, and <2 for other network statistics. If the *t*-ratio for

any P-Net configuration was over the threshold, we repeated the entire analysis by adding and/or

discarding configurations. All configurations from converged models are explained in Table 1

and all configurations in models are visualized in Figure 1.

Figure 1 Structural effects (configurations) included in converged ERGMs

Table 1 Interpretation of structural effects significant in converged ERGMs (Lusher et al., 2013)

ERGM structural effect	Interpretation
Reciprocity	The propensity for reciprocated tie formation. A positive effect indicates more than expected reciprocal collaboration.
2in-star, 3in-star	The propensity to receive ties (centralization around

	particular institutions). A positive effect indicates the presence of institutions important for network coordination.
2out-star, 3out-star	Tendencies for institutions to send ties. A positive effect indicates the presence of active institutions for spreading information.
2path (2mixed-stars)	Positive effect indicates the presence of bridging institutions (an organization with incoming ties tend to send a tie).
Alternating out-star	Positive effect indicates the presence of institutions that actively seek to transfer information (expansiveness).
Alternating in- and out-star	Assortativity (correlation) between in-degree and out-degree. A positive effect indicates a preference for an institution to send information to others that are similar in some way.
Alternating down and up k 2-paths	The propensity for two institutions to simultaneously send/obtain information from the same actors. A negative effect indicates a lower chance of exchanging information (competition among network members).

Institution-level centrality metrics

To analyze the position of institutions participating in grassland governance networks within the

three protected areas, we calculated the following node-level centrality metrics for one-mode

directed networks: normalized in-degree centrality, normalized out-degree centrality, eigenvector

centrality, and normalized betweenness centrality (Borgatti et al., 2017). In- and out-degree

centralities represent the number of ties an institution has divided by the maximum number of

possible ties in the respective set (incoming, outgoing). In-degree centrality measures the status

of the institutions (institutions with high in-degree are considered as hubs in resource

management, while those with high out-degree are important for initiating management activities

(Borgatti et al., 2017). Betweenness centrality for an institution is calculated as the number of

shortest paths that pass through the respective institution, normalized by the maximum

betweenness that any institution in the network can achieve. An institution with a high

betweenness score might play an important role in the network because it can mediate the

exchange of information between the linked institution (Hanneman and Riddle, 2005).

Eigenvector centrality of an institution, calculated as the principal eigenvector of the adjacency

matrix defining the network, and normalized by dividing the raw eigenvector score by maximum

score attainable in the network, capture not only the number of connections but also the influence

of neighbor institutions. Because it is a sum of the centrality values of the connected institutions,

eigenvector measure shows here how central is the institution of interest in the network (e.g., in

the best position to transfer information) (Borgatti et al., 2017). Network analyses were

performed using Ucinet 6.698 (Borgatti et al., 2002).

Perception of grassland related agri-environmental subsidies by farmers

To analyze how farmers from the three regions with past farming management perceive EU agri-

environmental measures, we interviewed farmers from the three protected areas were those

benefiting from one of the following CAP subsidy packages (Ministry of Agriculture and Rural

Development, 2014): a) CAP subsidy package 1 high nature value grassland, CAP subsidy

package 2 traditional agriculture (only in combination with package 1); b) CAP subsidy package

2 traditional practices; c) CAP subsidy package 3 grassland important for birds (*Crex crex*,

Lanius minor, *Falco vespertinus*). For package 1, the main requirements are indicated by the

interdiction to use pesticides and synthetic fertilizers, limited use of manure, delayed haymaking,

and low-intensity grazing. For package 2, the applicants must comply with requirements for

package 1 and opt for the manual mowing or using small machinery. Package 3 requires as a

supplement leaving 3 meters of untrimmed grassland (Sutcliffe et al., 2015). The survey

administrated to farmers includes capacity (owner, administrator), perception of CAP influence

on grassland management (no influence, negative influence, positive influence), and main issues

related to grassland management under CAP influence (insufficient funding per ha, late payment

of subsidies, bureaucracy, lack of market for grassland products). The questions were selected

after a face to face interviews with representatives of Agencies for Payments and Intervention for

Agriculture in the three protected areas. We collected 58 responses from farmers as follows: 9

from Macin Mountains, 24 from Sighisoara Tarnava Mare, and 25 from Iron Gates. The answers

were analyzed using Multiple Correspondence Analysis (MCA) (Greenacre and Blasius, 2006).

MCA allows the evaluation of associations among more than two categorical variables

(Greenacre, 1991). We performed MCA analyses using FactoMineR and factoextra R packages

(Kassambara and Mundt, 2016; Lê et al., 2008). We generated a symmetrically scaled biplot of

coordinates of coded answers (active variables) and included protected area and farmers' capacity

as supplementary variables (Greenacre and Blasius, 2006). We considered an answer as a

significant contributor to an axis if its eigenvalue exceeds the average of eigenvalues from the

respective axis (Kassambara and Mundt, 2016).

The results reported in this paper are available in GitHub

(https://github.com/rlaurentiu/grassland_networks).

Results

Network-level processes in grassland governance

The size of grassland governance networks varies between 63 (Iron Gates) and 31 organizations

(Macin Mountains) (Table 2). In all cases, public authorities and administrative bodies dominate

the network. In contrast, public and private enterprises and NGOs and research institutions are

only marginally presented (Supplementary Table 1). The grassland governance networks are

sparse in terms of densities of connections (Table 2). However, the network within Macin

Mountains is the most collaborative (22% of potential connections exist), followed by Iron Gates

(12% of potential connections exist) and Sighisoara (8% of connections exist). The three

networks are also dissimilar in terms of average connections between participating institutions

(average degree). Institutions involved in the grassland management in Iron Gates and Macin

Mountains are well linked when compared to Sighisoara (Kruskal-Wallis test in-degree = 23.89,

$p < 0.001$; Kruskal-Wallis test out-degree = 30.21, $p < 0.001$). Macin Mountains and Iron Gates

are also similar in terms of reciprocal connections, with over 80% of the ties reciprocated,

significantly larger than in Sighisoara, where 65% of connections between involved institutions

are in both directions. The average shortest path connecting two institutions (average geodesic

distance) is smaller in Macin Mountains than in Iron Gates and Sighisoara. A similar pattern

follows the network diameter, i.e., the number of institutions connecting the two most distant

connected institutions in the respective network or maximum geodesic distance.

Table 2 Cohesion and reciprocity metrics in three grassland governance networks from Romania

Network-level metric	Sighisoara	Iron Gates	Macin Mountains
Size	52	63	31
Average degree	3.98	7.33	6.58
Density	0.08	0.12	0.22
Fragmentation	0.15	0.12	0.12
Diameter	6	5	4
Average shortest path	2.66	2.31	1.92
Arc reciprocity	0.65	0.81	0.84

ERGMs with structural effects and fixed propensities for tie formation (network density)

converged for all analyzed networks and fitted very well the configurations of observed networks

(Table 3, Supplementary Table 2). ERGMs indicate a dissimilar pattern of interactions among

the network actors in the three analyzed networks. The propensity for collaborations (reciprocity)

is higher than expected by chance in Sighisoara and Macin Mountains but at random in Iron

Gates. The dissimilarity of Iron Gates can be explained by the positive 2-in-star effect, which

suggests a slightly higher than expected chance for an institution to receive ties from two others

(2-in-star centralization) but at random propensity of sending ties to two other institutions (2-out-

star popularity). The 2 in star effect, as well as the similar effect one order higher (3-out-star), is

positive in Sighisoara, suggesting the presence of popular actors. However, these popular actors

do not act as brokers in Sighisoara (receivers of ties also send ties) as 2-path configurations and

alternating out stars are lower than expected. Also, alternating out stars is significantly lower

than expected in Macin Mountains, suggesting the lack of coordination with more than one

institution. ERGM for Iron Gates network converged only by including a higher-order

configuration, which a lower chance to exchange information between several actors, confirming

the lack of reciprocity at higher orders.

Table 3 ERGM results for three grassland governance networks from Romania (estimates and

standard deviation, and t-ratio). Significant effects (p < 0.05) are in bold and marked with *.

Effects	Sighisoara		Iron Gates		Macin Mountains	
	Estimates (sd)	t-ratio	Estimates (sd)	t-ratio	Estimates (sd)	t-ratio
reciprocity	**4.35 (0.35)**	**-0.03***	-0.23 (0.26)	-0.04	**4.01 (0.42)**	**-0.03***
2-in-star	0.22 (0.14)	0.05	**0.13 (0.05)**	**-0.04***	0.35 (0.29)	-0.04
2-out-star	**0.68 (0.17)**	**0.03***	0.06 (0.06)	-0.03	0.16 (0.33)	-0.06
3-in-star	-0.003 (0.01)	0.03	-0.0003 (0.003)	-0.02	-0.01 (0.02)	-0.01
3-out-star	**-0.04 (0.02)**	**0.05***	0.004 (0.01)	-0.01	-0.003 (0.03)	-0.05
2path	**-0.16 (0.03)**	**-0.02***	0.01 (0.01)	-0.07	-0.05 (0.10)	-0.07
AinS	-0.97 (0.58)	0.01	-0.39 (0.74)	-0.07	-1.78 (1.20)	-0.07
AoutS	**-1.54 (0.55)**	**-0.02***	-0.36 (0.77)	-0.06	-1.63 (1.20)	-0.04
Ain1out-star	0.27 (0.50)	0.03	0.08 (0.08)	0.05	1.62 (1.14)	0.03

1inAout-star	0.82 (0.42)	0.04	-0.03 (0.06)	0.02	0.99 (1.21)	-0.01
AinAout-star	**-1.53 (0.60)**	**0.06***	-1.09 (0.84)	0.06	**-3.00 (1.06)**	**0.07***
A2P-DU	-	-	**-0.33 (0.04)**	**0.08***	-	-

Key stakeholders in grassland governance networks

The network of stakeholders involved in grassland management indicates that the most important actors mediating the transfer of information within the network (betweenness centrality) are mostly public agencies (Figures 2-4, Supplementary Table 2). In Sighisoara area, top institutions are county-level Agencies for Payments and Intervention for Agriculture (in charge with the management of application for subsidies), farmers and enterprises (as collective actors), and the representative of Ministry for Agriculture in Sibiu county. For Măcin, top stakeholders are the county level Agency for Payments and Intervention for Agriculture (in charge with the management of application for subsidies) and Office for Soil Studies (in charge with soil analyses and classification), followed by protected area administration and the local office of Agency for Payments. In Iron Gates, farmers as collective actors are the main actors mediating the transfer of information, followed by the administration of the protected area and local offices of the Agency for Payments. In Sighisoara and Macin Mountains networks, the number of actors with betweenness centrality of zero (actors with no role in the transfer of information) is higher than in Iron Gates.

Figure 2. Grassland collaboration network in Macin Mountains (green lines = reciprocal links of collaboration, red lines = one direction of collaboration, orange circles = public authorities, red circle = enterprises, yellow circle = public administration, green circle = NGO & research & education organizations, size of nodes = betweenness centrality). The full names of organizations are available in Supplementary Table 1.

Figure 3. Grassland collaboration network in the Sighisoara area (green lines = reciprocal links

of collaboration, red lines = one direction of collaboration, orange circles = public authorities,

red circle = enterprises, yellow circle = public administration, green circle = NGO & research &

education organizations, size of nodes = betweenness centrality). The full names of organizations

are available in Supplementary Table 1.

Figure 4. Grassland collaboration network in Iron Gates (green lines = reciprocal links of collaboration, red lines = one direction of collaboration, orange circles = public authorities, red circle = enterprises, yellow circle = public administration, green circle = NGO & research & education organizations, size of nodes = betweenness centrality). The full names of organizations are available in Supplementary Table 1.

When contrasting the in-degree (number of incoming collaborations) and the out-degree centralities (number of outgoing collaborations) with the eigenvector centrality (influential stakeholders in the network) (Figure 5, Supplementary Table 1), the top stakeholders are similar with those resulted from betweenness centrality, with the notable exception of two town halls in Iron Gates network (Moldova Noua and Garnic), involved in advising farmers to complying with the rules for subsidy payments.

Figure 5. In and out-degree versus eigenvector centrality in grassland collaboration networks in the three protected areas. The full names of organizations are available in Supplementary Table 1.

Farmer's perception of grassland governance

When analyzing the main issues of grassland governance as perceived by the farmers benefiting of agri-environmental subsidies resulted that the farmers of the Sighisoara area and Macin Mountains are slightly different from the farmers of Iron Gates on both dimensions of MCA biplot (Figure 6). When interpreting the answers with a contribution to a dimension above the average contributions of all variables in the respective dimension, it resulted that the farmers of the Sighisoara area and Macin Mountains consider that the CAP provides insufficient subsidies funds (dimension 1). At the same time, the farmers from Iron Gates consider CAP having a positive influence on grassland management, but there is an issue with late payments. On dimension 2 the pattern indicates that of Sighisoara and Macin consider CAP as providing a market for grassland products, while the farmers of Iron Gates consider CAP having a positive influence even if the payments are insufficient but too bureaucratic.

Figure 6. Biplot of MCA on farmer's perception of grassland governance (bold = significant on
dimension 1, italic = significant on dimension 2).

Discussion

By using network analyses and semi-structured interviews, we showed that coordination of
networks around grasslands management within protected areas with different past management
is influenced by the complexity of the administrative institutions at the local level and not by past
farming practices and land stewardship. Management actors from areas where the sense of
ownership is a long tradition (e.g., Sighisoara area, (Sutcliffe et al., 2013)) are not keen to
collaborate more than actors from areas with a centralized past, such as Macin Mountains. This

can be explained by the fact that Sighisoara area had a large ethnic and social change, with the

rapid collapse of the traditional Saxon society and institutions and the formation of new rural

communities in many of them being no Saxon ethnics anymore (Fischer et al., 2012; Hartel et al.,

2014). In this area, the current rural inhabitants often refer nostalgically to the Saxon society they

experienced as being dominated by strong social cohesion, well-defined rules, and communal

grassland and forest stewardship forms (Hartel et al., 2014). The political and institutional

instability of the recent decades (the communist and post-communist period) does not allow

forming of strong social and institutional capitals, the rural communities suffering from rigidity

and poverty traps (Mikulcak et al., 2015). Within the above context, the legacy of the past is

strong in the people's collective memory but does not help create a collaborative institutional

structure. Institutional networks in all three regions are coordinated by county-level public

authorities in charge of subsidy payments, which suggest that the EU's CAP did not genuinely

contribute to the formations of farmers' local associations. The willingness and capacity of

farmers to collaborate was not an explicit aim of the CAP, and previous research highlighted the

need for adjusting the CAP payment criteria to foster collaborations at local levels (Leventon et

al., 2017).

The number of management actors involved in grassland management is larger in Iron Gates and

Sighisoara area than in Macin Mountains. The likely reason for this finding is that the region is

part of more counties (two in Iron Gates, three in the Sighisoara area compared to one county in

Muntii Macinului). Local authorities and public administration dominate collaboration networks

in all three regions (84% of actors in Macin Mountains, 81% in the Sighisoara area, and 77% in

Iron Gates); thus, the size of the network was influenced by administrative jurisdiction and not

by diversity of non-public actors. The importance of inclusion of the local, informal institutions

and knowledge types into the management of multifunctional landscapes with high socio-cultural

and natural values was suggested as a key element for leveraging sustainability transformations

(Balázsi et al., 2019). The results of our analyses also reveal a strong relationship between the

farmers and the conventional institutional structures involved in the grasslands governance and a

weak influence of the protected area management institutions on their grassland governance

practices (see Table 2).

The effect of territorial jurisdiction is also evident when analyzing the mean number of

connections of an institution. Institutions participating in grassland management in the

Sighisoara area are the least connected (3.98 connection per institution) while in Iron Gates, and

Macinului Mountains are well connected (average degree = 7.33 in Iron Gates and 6.58 in Macin

Mountains). The actors from the Sighisoara area also display lower reciprocity than the two other

networks (65% comparing to over 80%, Table 2). This suggests low levels of bidirectional

collaborations between the grassland governance institutions, implying a unidirectional flow of

information and power influence within the network. This could represent a challenge in the

front of sustainability initiatives, which also implies that the existing grassland governance

network is capable to 'absorb' genuine information from the outside and turn it into opportunities

of grassland management. Indeed, several research initiatives and research communication

activities targeting ancient grasslands exist in the Sighisoara region (Hartel et al., 2016);

however, virtually no information turns into a real opportunity for the sustainability of these

ecosystems.

When interpreting the ERGM results (see Table 3 and SI), the historical management effects on

the patterns of relationships in the three protected areas are hindered by the top-down

prescriptions of the CAP. The tendency for collaboration between actors is higher than expected

in the Sighisoara area and Macin Mountains and without a pattern (higher/lower than expected in

Iron Gates (reciprocity effect Table 3). The reciprocity of collaboration in the Sighisoara area is

reduced when compared with Iron Gates and Macin Mountains (only 65% compared to 85% in

Macin Mountains). However, because the reciprocity is stronger than expected at network-scale

effect, there is a high potential for improving if non-public organizations and protected area

manager will intensify their trans-counties collaborations (for example, a more active local agro-

economy). On the contrary, the higher than expected reciprocity from Macin Mountain suggests

that the only way to improve the collaboration is by involving other actors in the network than

existing ones (e.g., farmers associations, small enterprises).

In the Sighisoara area, the governance network shows a tendency for sending a request for

collaborations to only two other organizations (2-out-star effect positive). However, the effect is

reversed at higher orders (e.g., sending a request for collaborations to only three other

organizations, 3-out-star negative). The effects for incoming ties (i.e., 2-in-star, 3-out-star) are at

random in the Sighisoara area and Macin Mountain, suggesting the lack of popular actors (actors

receiving more collaboration requests than expected). Such actors are present in Iron Gates (2-in-

star positive), suggesting that high ranking actors in terms of centrality metrics are network

leaders.

In the Sighisoara area, the management network also shows a lack of bridging organizations (2-

path negative) and a substantial lack of appetence to transfer information to other organizations

in the network (AoutS negative). The lack of institutions actively seeking to expand the

connections is coupled with avoidance of collaboration between institutions with a similar level

of activity (lower or higher) in terms of in-degree and out-degree (AinAoutS negative) complete

the picture of an unexpected low collaborative network within Sighisoara area. In the two other

networks, most effects in the converged models are at random, with the notable exception of

reciprocity (higher than expected) and avoidance of collaboration between institutions with a

similar level of activity (AinAoutS negative) and a low degree centralization around popular

institutions (2-is-star positive) and a lower chance to break this centralization (A2P-DU negative)

in Iron Gates.

The key stakeholders by contribution to information flow in the network (network members with

high betweenness centrality, see Figure 2-4) are local Agencies for Payments and Intervention

for Agriculture in the Sighisoara area and Macin Mountains, and farmers as collective

institutions and Iron Gates Natural Park Administration in Iron Gates. From this point of view,

CAP contributed to bureaucratic governance in the Sighisoara area (see also (Mikulcak et al.,

2013)) and Macin Mountains but to a lesser degree in Iron Gates, where the protected area

administration is very active. The same pattern is evident when considering eigenvector

centrality (higher eigenvector suggesting a higher power of institutions participating in the

respective network) and in-degree and out-degree (activity in the network) (see Figure 5). In the

Sighisoara area and Macin Mountains, the agencies distributing the CAP subsidies are most

active and influential, while in Iron Gates, some farmers, the protected area administrator, and

city halls dominate the network.

The history of the investigated protected areas fundamentally influences the farmers and their

perceptions and actions on the grassland management (Hanspach et al., 2014); however, this

study suggests a diminished influence after only a few years of CAP subsidies. The MCA results

(see Figure 6) indicate that the farmers from the Sighisoara area and Macin Mountains perceive

the CAP differently than farmers from Iron Gates. Farmers from the Sighisoara area and Macin

Mountains have no convergent opinion about CAP influence, while the farmers from Iron Gates

(mostly small owners) recognize a positive influence of the CAP. This pattern might be because

grasslands from the Sighisoara area and Macin Mountains are lucrative business while in Iron

Gates are mostly used for subsistence (Manea, 2003; Otiman-Păun, 2006; Rey et al., 2007), thus,

receiving money from CAP subsidies is considered by Iron Gates farmers as a good addition to

the local economy. The need for additional income s confirmed by the complaining about late

payments and insufficient funds. Late payments are not an issue for farmers from the Sighisoara

area and Macin Mountains (mostly administrators) but insufficient funds. Notably, in the

Sighisoara area and Macin Mountains, farmers consider CAP as providing a market for their

products, which is in part because of land concentration in these areas (Calota and Patru-

Stupariu, 2019).

Traditionally managed grasslands suffer threats such as intensive, highly specialized

management of grasslands, land abandonment (Dorresteijn et al., 2013; Hartel and Plieninger,

2014). Although farmers receive subsidies for non-intensive management of their grasslands,

they do not consider this as an aid to bring prosperity to rural biodiversity and cultural heritage

conservation (Öllerer, 2013). Our results showed that in the investigated case studies, CAP and

protected area regulations failed to reverse the threats to traditionally managed grasslands and

did not foster genuinely new institutional collaborations. Farmers from former state-owned areas

are still looking for state advice, while those from areas with a liberal past are not keen to

collaborate, and the networks are centralized around regional and local public bodies.

Conclusions

The main obstacles for effective collaborations for the multifunctional management of the

grasslands identified by the farmers are limited financial incentives, lack of market for their

products, absence of consultants, and lack of follow-up considering the effectiveness of the

measures in protecting the high-value biodiversity. This research avenue significantly contributes

to the consolidation of using social network analysis for understanding multifunctional landscape

governance worldwide (Barnes et al., 2017). Consequently, our results provide a solid base for

further applications of social science in ecology and conservation biology and will identify key

milestones for understanding the evolution of coupled social-ecological systems (Bodin et al.,

2019). From the ecological and social point of view, we identified the barriers to the

implementation of CAP through the analysis of relevant landscape governance networks in

Romania and determined how network structures influence efficacy in maintaining biodiversity.

At the European level, the strengthening of the grassland management advisory system needs to

be deepened, and each strategic plan of the CAP should include a section on how to stimulate

knowledge exchange and innovation.

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