

## SPATIAL DRIVERS OF AGROFORESTRY ADOPTION SUPPORTED BY CAP MEASURES IN HUNGARY



K A P O S V Á R I

E G Y E T E M

*Arnold Csonka*

*Tibor Bareith*

*Veronika Gál*

*Imre Fertő*

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# Introduction

*Agroforestry (AF) systems are complex agricultural systems that can combine multiple land use in time and space.*

*It has implemented multiple uses of the area.*

(Saláta et al.. 2013. p. 315)

*Main principle: **exploiting synergic ecological and economic interactions.***

(Lungren és Raintree. 1982; Leakey. 1996; Nyári. 2006).



1 ha agroforestry (silvoarable) land =  
0.8 ha arable land + 0.6 ha forest  
(Dupraz. 2012)

**Generating externalities:**

- **Biodiversity**
- Varied landscapes
- Deflation, **erosion prevention**
- **Carbon absorption**
- **Activity diversification**

(Keserű és Honfy. 2015)

# Motivation and objective 1 – policy issues

- Most important AF promoting CAP-measures budgeted in Hungary between 2007 and 2013 (-2015\*):
  - Measure 221: First afforestation of agricultural land
  - Measure 222: First establishment of AF systems on agricultural land
- Very low implementation ratio:
  - Measure 221: 58% implemented to the amount of the money budgeted
  - Measure 222: 26% implemented to the amount of the money budgeted
- Future policy challenges (Mosquera-Losada et al, 2017):
  - CAP should promote AF through payments to enhance AF practices
  - European AF Strategy should be designed to foster AF in Europe



**There is a need for a better understanding of drivers affecting the uptake of AF practices**

# Motivation and objective 2 – research questions

- There is a relative low number of scientific studies regarding environmental socio-economic drivers of AF adoption.
- The vast majority of these studies investigates the environmental factors at spatial level OR socio-economic factors at farm(er)-level.
- Spatial analysis dealing with both environmental and socio-economic factors of AF adaptation is virtually non-existent.

***The aim of the research is to explore the spatial hotspots of AF adoption supported by CAP and identify their environmental and socio-economic drivers at settlement level.***

- RQ1: Where can the hotspots of AF adaptation be found in Hungary?
- RQ2: What are the most important drivers shaping the spatial clusters of AF adoption?

# Theoretical model of the research

Drivers of AF adoption			
Settlement size	<i>Population</i>		<i>Forests</i>
	<p><b>Proxy variable for Spatial pattern of AF adoption:</b></p> <p>Density of subsidies (HUF/ha farmland) spent in measures 221 and 222 at settlement level (2007-2015)</p> <p>Variable code: <i>suppaftot</i></p>		<p><i>Grasslands</i></p> <p><i>of natural areas</i></p> <p><i>farm areas</i></p> <p><i>forests</i></p> <p><i>Grasslands</i></p> <p><i>Croplands</i></p> <p><i>from the district</i></p>
Financial support for sustainable technologies			
Farm animals			
Farm structure	<i>Ratio of individual farms</i>	<b>cooperations</b>	<i>Financial support for producer groups</i>

# List of independent variables

<b>Variable code</b>	<b>Variable name</b>	<b>Period</b>	<b>Dim.</b>
<i>pop</i>	Settlement population	2011	heads
<i>totarea</i>	Total area (TA) of the settlement	2011	ha
<i>forest</i>	Ratio of forests to TA	2012	%
<i>natgrass</i>	Ratio of natural grasslands to TA	2012	%
<i>wetland</i>	Ratio of wetlands to TA	2012	%
<i>protarea</i>	Ratio of protected areas to TA	2012	%
<i>farmarea</i>	Ratio of farmlands (FL) to TA	2010	%
<i>agriforest</i>	Ratio of on-farm forests to FL	2010	%
<i>agrigrass</i>	Ratio of on-farm grasslands to FL	2010	%
<i>Istock</i>	Livestock density	2010	LU/ha
<i>rumin</i>	Ratio of ruminants to livestock pop.	2010	%
<i>empagri</i>	Ratio of agriculture in employment	2011	%
<i>indfarm</i>	Ratio of individual farms	2010	%
<i>agricoop</i>	Number of agricultural cooperatives	2007-2015 (avg)	pcs
<i>supppg</i>	Financial support for producer groups to FL	2007-2015 (sum)	HUF/ha
<i>suppextser</i>	Financial support for agricultural extension to FL	2007-2015 (sum)	HUF/ha
<i>suppagrenv</i>	Financial support for agri-environmental projects to FL	2007-2015 (sum)	HUF/ha
<i>distdc</i>	Distance from the district center	2011	minutes

# Methodology of empirical analysis

Natural logarithmization of ,suppaftot' variable  
→ ,log\_suppaftot'

Spatial Cluster Analysis of ,log\_suppaftot'  
global Moran's I and Getis-Ord  $G_i^*$

Save Spatial Cluster membership as a binary dependent variable: ( $c\_id=1$  if a settlement is the member of a high cluster,  $c\_id=0$  otherwise)

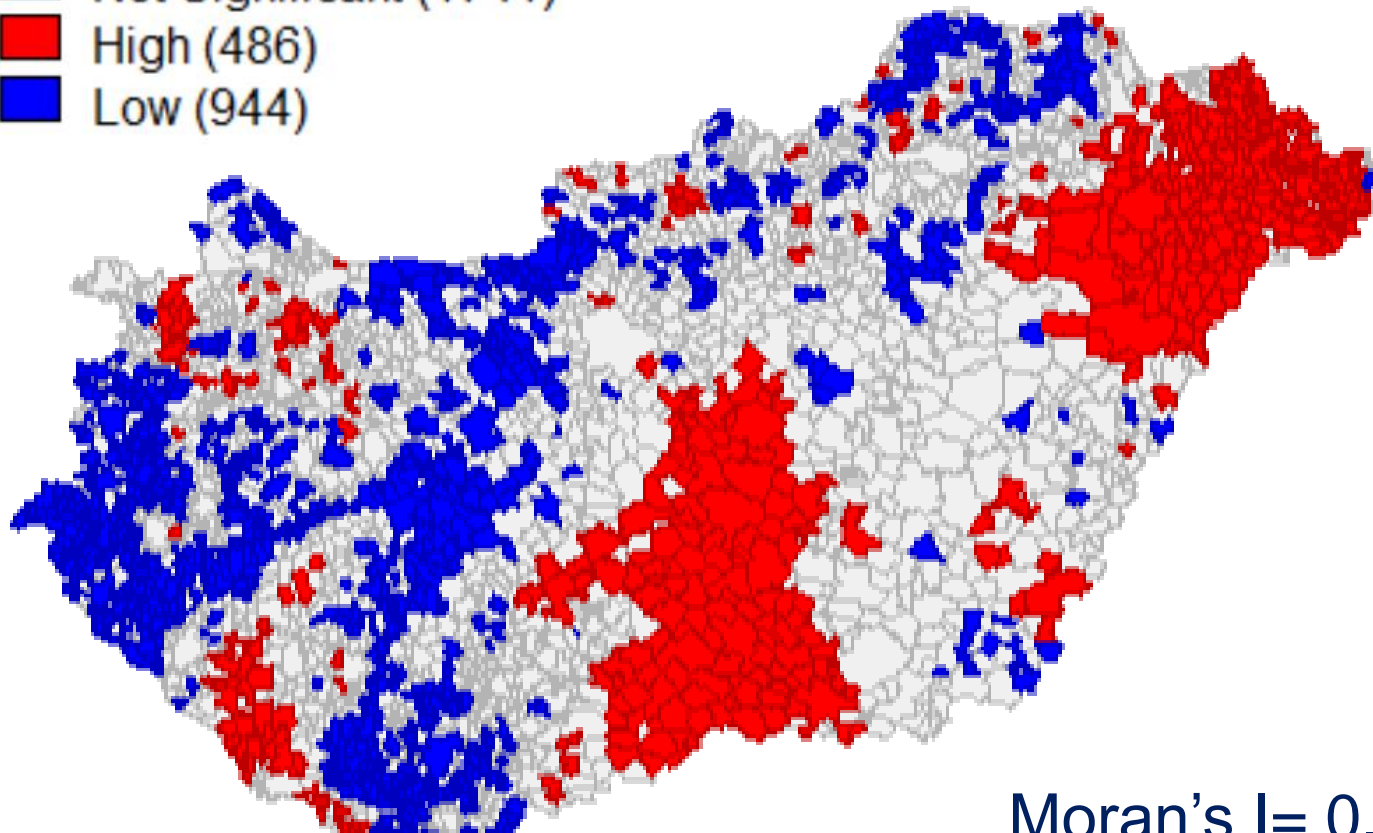
Factor analysis of the independent variables  
(Principal component. Varimax rotation)  
KMO. Bartlett test. Eigenvalue > 1

Logit regression using  $c\_id$  as binary dependent variable  
Goodness-of-fit and correctly classified index

# Results 1: Spatial Cluster Analysis

## $G_i^*$ cluster map of dependent variable

- Not Significant (1741)
- High (486)
- Low (944)



Moran's  $I = 0.30$   
Pseudo  $p < 0.05$   
Permut. = 10 000

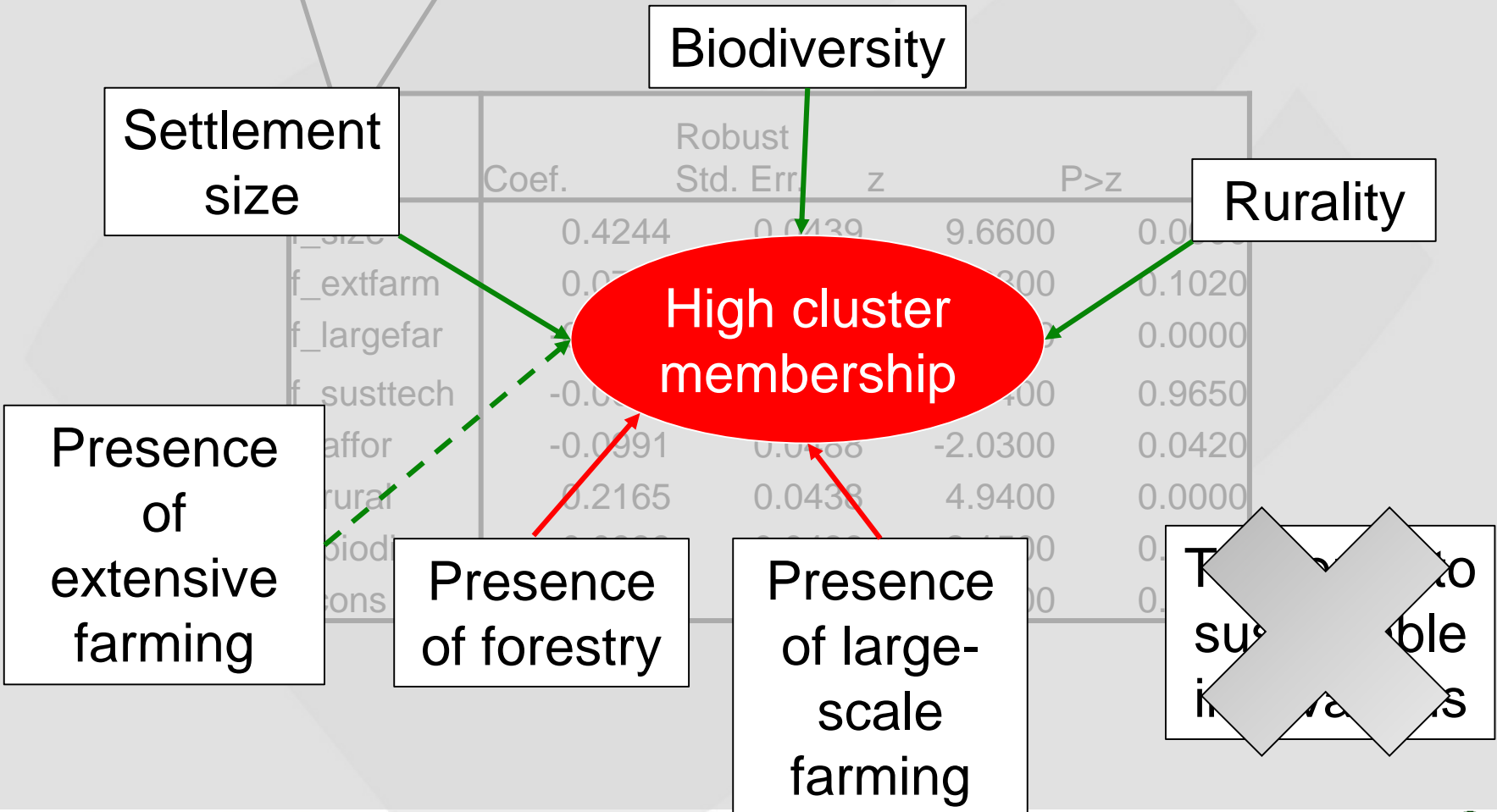


# Results 2: Factor analysis

Variable	Settlement size F_Size	Presence of extensive farming F_ExtFarm	Presence of large-scale farming F_LargeFarm	Tendency to sustainable innovations F_SustTech	Presence of forestry F_Affor	Rurality F_Rural	Biodiversity F_Biodiv
TotArea	0.8660						
Lstock	0.8275						
Pop	0.7034					- 0.4548	
AgriGrass		0.9103					
AgriCrop		- 0.8281					
Rumin		0.5776					
IndFarm			- 0.8136				
FarmArea			0.7565		- 0.3352		
AgriCoop	0.3969		0.4972				
SuppExtServ				0.9587			
SuppAgrEnv				0.9511			
SuppPG							
Forest					0.7559		
AgriForests					0.7360		
EmpAgri						0.7989	
DistDC						0.6749	
Wetland	<div style="border: 1px solid black; padding: 5px; display: inline-block;">                     KMO= 0.6393                      Barlett p-value&lt; 0.001                 </div>						0.7725
ProtArea						0.3849	
Natgrass					0.3354		0.4556

# Results3: Logit regression

0=low cluster or not significant  
1=high cluster



Goodness of fit:  $\chi^2(3129)=3110.7$   $p= 0.5884$  ✓  
 Correctly classified= 84.43%

# Conclusions 1

- The **structure of spatial drivers** identified in this research **partly differs from the farm-level models**. Our results contribute to developing an empirical tested model explaining spatial diffusion of AF practices.
- According to recent studies the **rural character of small-scale farming and the presence of extensive farming affect AF adoption positively**. The **positive role of settlement size** is probably ensured by the **access of markets and institutions**.
- Based on conceptual and empirical literature, we assumed that the high level of afforestation has a positive impact on AF adoption. **Surprisingly**, our results show that the **presence of traditional forestry is a barrier to apply AF practices**. Due to this fact the role of forests should be distinguished from other areas of high biodiversity (grasslands and wetlands).

# Conclusions 2

- Contrary to the recent farm-level surveys, **our results do not confirm that the tendencies to submit other agri-environmental applications and to use of extension services have a positive impact on application of AF practices.**
- Our results suggest that the wider adoption of AF technologies **needs a more consistent AF strategy and policy.** Future CAP should promote AF both on agricultural lands and in forests. The low AF **activity of highly forested areas can be improved by the promotion of forest farming and mountain-linked AF practices (i.e. mountain pastoralism).**
- **The connection between AF measures and other agri-environmental supports should be improved in the next CAP-period.**
- Further possible research direction: Our next aim is to develop and test a spatial weighted autoregression model.

**Thank You for Your  
Attention!**