

# Hybrid Agent-based Modeling of Land-use Changes in the Soyang River Basin

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Many factors such as climate change, new regulations, or local development plans can affect land-use decision-making processes in the Soyang River Basin. However, it is difficult to represent these factors in a model and to reflect the complexity of decision-making processes of land-users (Parker et al., 2003). In this situation, agent-based modeling (ABM) approaches can be useful to consider such diverse factors and emergent aspects in agent decision-making processes. Simulation of land use and cover changes based on the ABM approach can be effective to prepare land use and cover scenarios for support of future policy-making (Le, et al., 2008).

# **Research Objective** To develop a model that provides insight into critical agents' land-use decision-making and the emergent community-landscape transitions that result from policy change

#### **Research Methods**

- Study Area: Soyang River Basin area, South Korea
- Data: Spatial data, statistical data, spatially-explicit survey data
- Methods: Interviews and surveys with stakeholders, logistic regressions, NetLogo programming of ABM (including nested cellular automata)

#### Model Description and Procedures



## **Preliminary Results**

A. Patterns of land-use changes

	Land-Cover 2000						
		Other	Urban	Forest	Agri-culture	Sum	
Land-cover 1990	Other	1730 (25.45%)	718 (10.56%)	2134 (31.39%)	2216 (32.60%)	6798	
	Urban	170 (4.66%)	1766 (48.38%)	473 (12.96%)	1241 (34.00%)	3650	
	Forest	1510 (0.22%)	717 (0.11%)	665242 (98.34%)	8970 (1.33%)	676439	
	Agri-culture	1799 (5.25%)	2276 (6.64%)	11561 (33.73%)	18641 (54.38%)	34277	
	Sum	5209	5477	679410	31068		
▲ Change Matrix of Land-Cover							
	Urban(%)		Forest(%)			Agriculture(%)	
10							



B. Factors of land-cover changes and parameterization of cellular automata sub-model

Nodel:      Model:      Model:<	Model3: Agriculture to Forest Coefficie P-value -007 000
Model:      Model2:      Model2:      Model2:      Model2:      Model2:      Variable      Variable      Model2:      Model2:      Variable      Variable      Model3:	Agriculture to Forest Coefficie nt (B) P-value
Variable      Coefficie nr (B)      Pushue      Coefficie nr (B)      Pushue      Coefficie carificii      Pushue      Coefficie pushue      Pushue      Pushue <t< td=""><td>Je Coefficie P-value</td></t<>	Je Coefficie P-value
Rainfall .002 .000** .008 .000** .002 .000** Rainfall .002 .001**002 .021*001 .059 Rainfall .002 .001** .000	- 007 000
Elevation002 .000**001 .004**002 .000** Elevation003 .000** 000 311003 .000** Elevation003 .000** -001 .0	**000. 000.
Aspect .000 .744 .000 .096 .001 .000** Aspect .001 .001* .001 .001* .001 .001* .001 .001	**000. 000.
Slope025 .000**032 .000**061 .000**061 .000**01 .001 .001 .002 .000 .001 .000 Slope .009 .000**021 .0	.024 .000**
Upslope .048 .527 .292 .000** .042 .399 Slope051 .000**064 .000**044 .000** Upslope .074 .001** .046	.100 .000**
Wetness .019 .001** .007 .141 -011 .005** Upslope .203 .000**041 .660 .121 .000** Wetness .008 .001**018 .0	.007 .000**
D_Stream .001 .000** .000 .448 .000 .000 Wetness016 .058014 .227017 .001** D-Urban .000 .000002 .0	.001 .000
N_Other .024 .000** .018 .000** .009 .001** D_Stream .001 .000**002 .000**001 .000** D Stream .001 .000**001 .00	.001 .000**
N_Urban -024 .000** -018 .000** -020 .000** N_Forest -055 .000** -062 .000** -040 .000** N_Forest -012 .000** -020 (	
N_Forest .027 .000 .058 .000** .033 .000** N_Farm015 .000**011 .006** .007 .000** N_Farm020 .000** .025	002 001**
N_Farm -014 .000** .024 .000** .023 .000** R_park=0503 .002 -1.158 .000**278 .003 P_park=0658 .000**144	119 070
R_park=0378 .327856 .005**535 .071	
R_park=1 R_p	
R_water=0 .628 .319996 .079585 .040 R_water=0150 .050 -1.004 .013064 .012 R_water=0268 .036318	.000 .569
R_water=1 R_water=1 R_water=1	_
Constant -4.391 .000 -14.340 .000 -3.250 .000 Constant 4.779 .000 6.078 .000 3.953 .000 Constant -3.478 .000** .239	-10.940 .000

▲ Temporal Land-use Changes

#### **Expected Outputs**

- ① Calibrated cellular automata sub-model of land cover (broad type) transition at watershed level
- ② Calibrated sub-model of agents' decisions and their effect on agricultural land use
- **③** Landscape agent-based model that couples (1) and (2)
- Informed scenarios on future land-use change across the study watershed

### Contribution to TERRECO

- Reliable land-use maps to allow estimation of future potential changes in ecological processes and ecosystem services
  Definition of according to allow estimation of future potential changes in action and allowed allowed
- Definition of scenarios for land-use change in relation to possible changes in policy and climate

Reference Le, Q.B., Park, S.J., Vlek, P. L.G., and Cremers, A. B., 2008, Land-Use Dynamic Simulator(LUDAS): A multi-agentsystem model for simulating spatio-temporal dynamics of coupled human-landscape system. I. Structure and theoretical specification, *Ecological Informatics*, 3, 135-153. Parker, D.C., Manson, S.M., Janssen, M.A., Hoffmann, M.J., and Deadman, P., 2003, Multi-AgentSystems for the Simulation of Land-Use and Land-Cover Change: A Review, *Annals of the Association of American Geographers*, 39(2), 315-337