10년 타워 플럭스 관측에 기초한 농업생태계의 생태학적 지표 평가

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Assessment of Ecological Indicators of Agricultural Ecosystem Based on a Decade-Long Tower Flux Measurement

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I. Introduction

Ecological indicators (EI) are developed based on the framework on ecosystem structure and function, which are constrained by the flows of energy, matter and information. Nielsen and Jørgensen (2013) have identified three major directions in the development of EI: 1) biotic (i.e. related to already well-known and well-established classical indices in ecology), 2) network (i.e. based on various directions of network theory) and 3) thermodynamic (i.e. mainly derived from physics either first or second law of thermodynamics).

Field observation of micrometeorology including eddy covariance (EC) flux measurement provides the quantitative assessment of energy, matter and information flows in ecosystems. EC measurement has advantages for developing EI by offering continuous and long-term time series data for various variables with wide ranges of environmental conditions, along with the availability of global network with open access data (Baldocchi *et al.*, 2001). By employing the information theory to such time series, EC measurement can also be used for the assessment of biotic, network and thermodynamic indicators, which are available for the same system both spatially and temporally.

In this study, we focused on assessing the biotic and thermodynamic EI derived from EC measurement in agricultural ecosystem. In this study, the biotic indicators which are derived from many traditional measures include net ecosystem exchange (*NEE*), gross primary productivity (*GPP*), crop coefficient (K_c), and water use efficiency (*WUE*)). Thermodynamic indicators used in this study are based on entropy balance (dS/dt) (Brunsell *et al.*, 2011) as well as energy capture (R_n/R_{snet}) and

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dissipation ability (in terms of thermal response number, TRN) (Kutsch *et al.*, 2001; Lin *et al.*, 2009; Lin *et al.*, 2011). We expected that the integration of biotic and thermodynamic indicators will provide better holistic representation of the system state of the agricultural ecosystem.

II. Materials and Methods

The EC measurements of CO_2 , water and energy at Haenam Farmland in Korea (HFK) site over rice growing season from 2003 to 2012 were used in this study. Flux data processing was conducted using the KoFlux data processing protocol (Kwon *et al.*, 2009). The biotic and the thermodynamic EIs used in this study are presented in Table 1.

No	Category	variables	Symbol	Unit		
		Net ecosystem exchange	NEE	g C m ⁻²		
	Biotic indicator	Gross primary productivity	GPP	g C m ⁻²		
		Ecosystem respiration	RE	g C m ⁻²		
1		Evapotranspiration per precipitation	ET/P	unitless		
		Bowen ratio	b	unitless		
		Crop coefficient	K_c	unitless		
		Water use efficiency	WUE	g C kg H_2O^{-1} hPa		
2	Thermodynamic indicator	entropy balance	ds/dt	MJ m ⁻² K ⁻¹		
		energy capture	R_n/R_{snet}	Unitless		
		energy dissipation	TRN	$MJ m^{-2} K^{-1}$		

Table 1. Ecological indicators tested for agricultural ecosystem in this study

III. Results

3.1. Biotic indicators

In terms of carbon exchange during the rice growing season (Table 2), the averaged *NEE* during the study period was -113 ± 56 g C m⁻² with the peak carbon uptake in 2008 (-176 g C m⁻²) and the lowest uptake in 2012 (-4 g C m⁻²). From 2004 to 2009, the agricultural system remained a strong carbon sink. Then, from 2010, the sink strength became weaker. The averaged *GPP* during the rice growing season was 838 ± 41 g C m⁻², amounting up to approximately 70% of the annual *GPP*. The *GPP* varied with a minimum in 2003 (782 g C m⁻²) and a maximum in 2006 (901 g C

m⁻²). The *RE* averaged to be 726 \pm 48 g C m⁻² (about 64% of the annual *RE*) and fluctuatued with a tendency to increase toward the end of period.

ET during the rice growing season was 375 ± 24 mm, accounting for ~60% of the annual total. The ratio of *ET* to *P* was on average 0.41 ± 0.08. The *ET* in 2008 (driest year) accounted for 57% of *P* while only 33% in 2003 (wettest year). The averages of *H* and *LE* were 240 ± 19 and 914 ± 57 MJ m⁻², respectively. Hence, the *b* (= *H/LE*) was on average 0.26 ± 0.03 with the highest in 2008 and the lowest not in 2003 but in 2012.

In terms of water use, the growing season average of K_c was 0.94 \pm 0.07. The K_c values fluctuated with a maximum of 1.04 in 2012. On the other hand, the *WUE* was on average 22.25 \pm 3.37 g C kg H₂O⁻¹ hPa. From 2004 to 2009, *WUE* was higher than the average and then lower thereafter.

N	Category	EI	2003	2004	2006	2008	2009	2010	2011	2012	AVG	std
1	Biotic indicator	NEE	-88	-165	-148	-176	-160	-91	-68	-4	-113	56
		GPP	782	851	901	890	806	866	803	808	838	42
		Re	694	685	753	714	646	775	735	803	726	48
		ET/P	0.33	0.32	0.41	0.57	0.40	0.48	0.39	0.36	0.41	0.08
		b	0.26	0.25	0.32	0.26	0.30	0.24	0.27	0.21	0.26	0.03
		Kc	0.78	0.89	0.97	0.95	0.94	0.97	0.97	1.04	0.94	0.07
		WUE	17.9	26.6	25.3	25.5	24.8	19.4	19.8	18.7	22.3	3.4
2	Thermodynamic indicator	dS/ <i>dt</i>	1.11	1.09	1.02	0.76	1.12	0.92	0.88	0.83	0.97	0.13
		R _n /R _{snet}	0.78	0.75	0.72	0.72	0.72	0.76	0.75	0.75	0.74	0.02
		TRN	0.96	0.90	0.82	0.74	0.76	0.88	0.78	0.84	0.84	0.07

Table 2. Ecological indicators over rice growing seasons at HFK

Unit: NEE, GPP, $Re = g C m^{-2}$, ET/P, b, K_c , $R_n/R_{snet} =$ unitless, WUE= g C kg H2O⁻¹ hPa, ds/dt, TRN= MJ m⁻² K⁻¹.

3.2. Thermodynamic indicators

The changes in entropy with time (dS/dt) were positive with an average of 0.97 ± 0.13 MJ m⁻² K⁻¹, indicating the overproduction of entropy in this agricultural ecosystem. In general, however, decreased from 2003 to 2012 except a sudden drop in 2008 and the recovery in 2009, thereby gradually approaching the dynamic equilibrium.

In terms of energy capture, R_n/R_{snet} was on average 0.74 ± 0.02, which was higher than the annual R_n/R_{snet} (i.e., 0.59 ± 0.03). The measure of energy dissipation, *TRN* was on average 0.84 ±

0.07 MJ m⁻² K⁻¹, higher than the annual *TRN* (0.54 \pm 0.05). During the rice growing season, the enhanced energy capture resulted in more energy dissipation, which also lowered the gradient of surface temperature.

3.3 Integration of biotic and thermodynamic indicators

It is important to integrate and summarize the multiple EIs in a way that not only experts but also stakeholders can understand their meanings. By providing such an integration, the users of these EIs can easily understand the behaviors of the indicators against some conditions (e.g. disturbances). In Fig. 1, we used the amoeba diagram method to synthesize the EIs by comparing and contrasting the two different cases: when EI was higher than the average and when EI was lower than the average of the representative biotic and thermodynamic indicators (i.e., *NEE* and dS/dt).

Based on NEE (Fig. 1a), for the period when the agricultural system absorbed more carbon (i.e., higher *NEE*) than the average, we note higher *WUE*, higher β , and higher *ET/P*, while other EIs showed no significant differences.



Fig. 1. Amoeba graphs of the EIs based on the contrasting conditions of (a) NEE and (b) dS/dt.

 K_c and RE. Relatively insignificant changes in GPP, R_n/R_{snet} and TRN suggest that these indicators were not the causes of the enhanced NEE and WUE. The lack of sensitivity of R_n/R_{snet} and TRN to changes in dS/dt suggests that these two thermodynamic indicators may be good indicators for self-organization but may be inadequate for holistic EIs. Our results provide further implication that the triple wins (i.e., more production, less carbon emission, and better resilience) pursued by climate smart agriculture (CSA) would be a difficult challenge facing the CSA communities.

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