



Joint conference of AsiaFlux Workshop 2017
and the 15th Anniversary Celebration of ChinaFLUX

August 17-19, 2017 Beijing, China



The spatial variability of water use efficiency and its underlying mechanisms in terrestrial ecosystems of China

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2017.08.17 Beijing China

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Outlines



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Introduction

2

How WUE spatially varied?

3

Why WUE spatially varied?

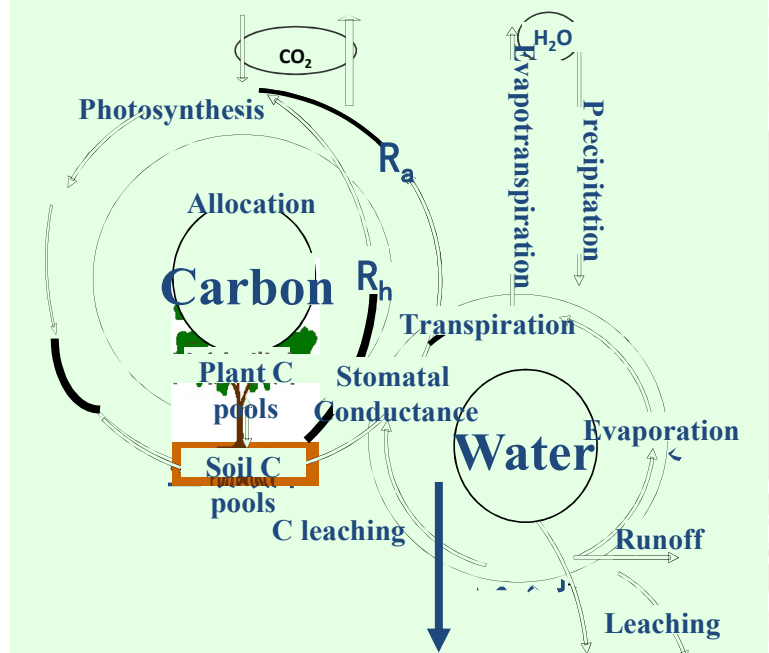
4

Summary



Introduction

Budget of C and H₂O fluxes in ecosystems



Dynamics of C source/sink in ecosystems

Coupling relation between C and H₂O cycles

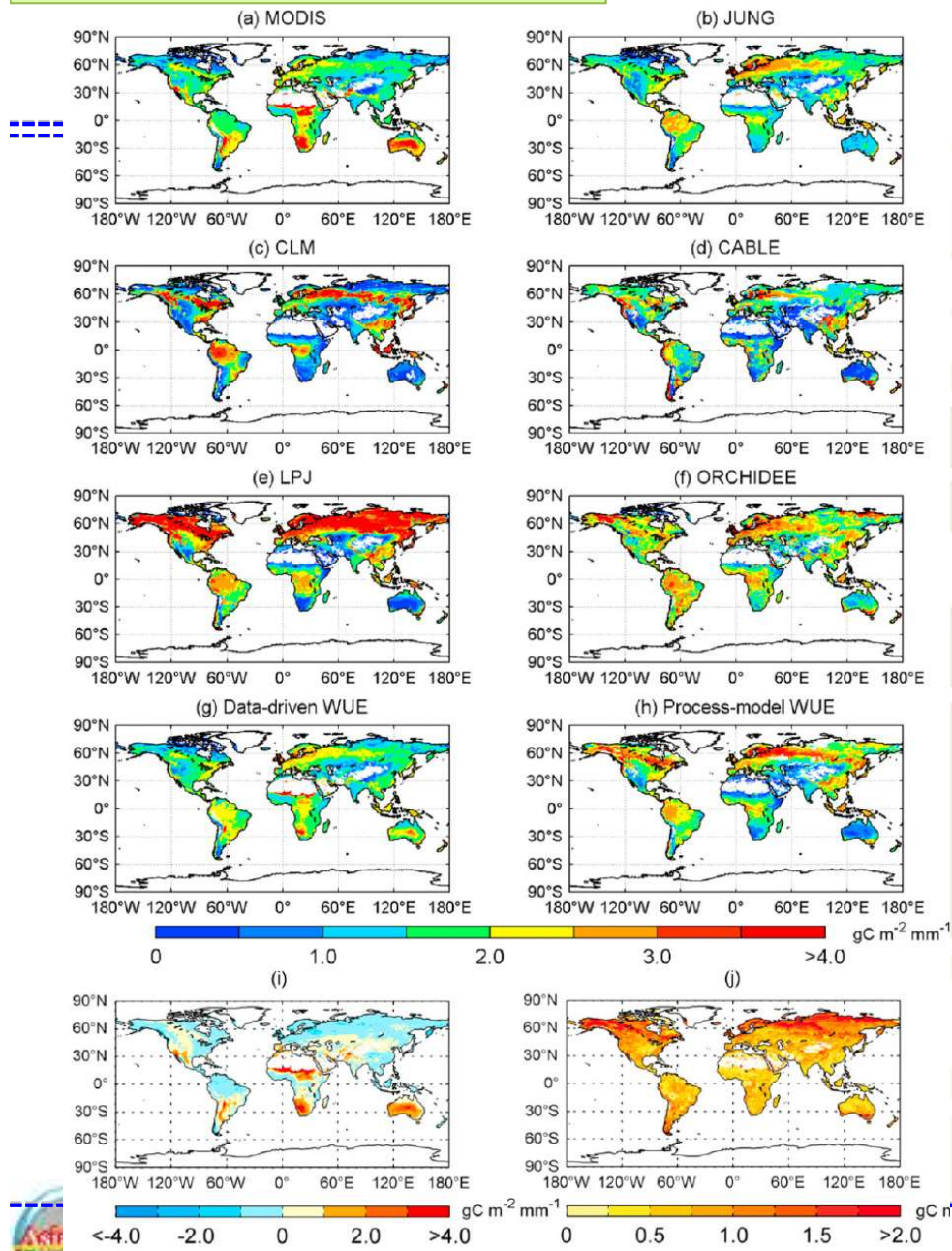
spatial pattern of climate

water cost of carbon uptake

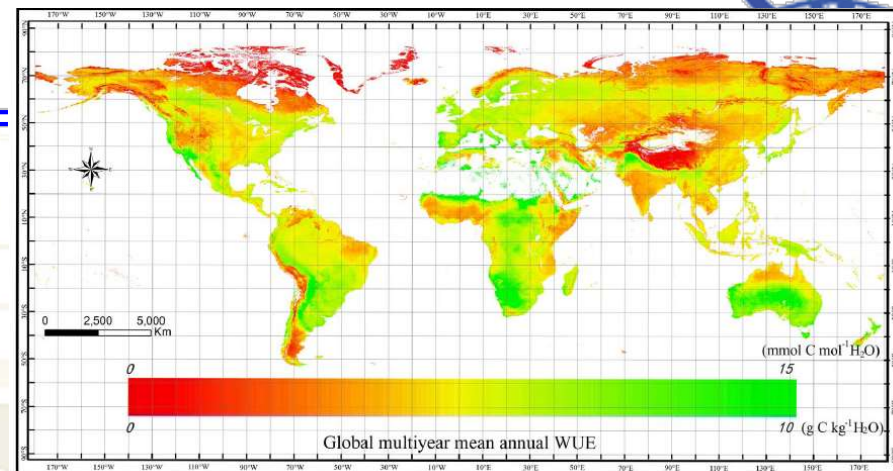
WUE spatial variability

Understanding on carbon cycle process at a large scale

Spatial variation of WUE



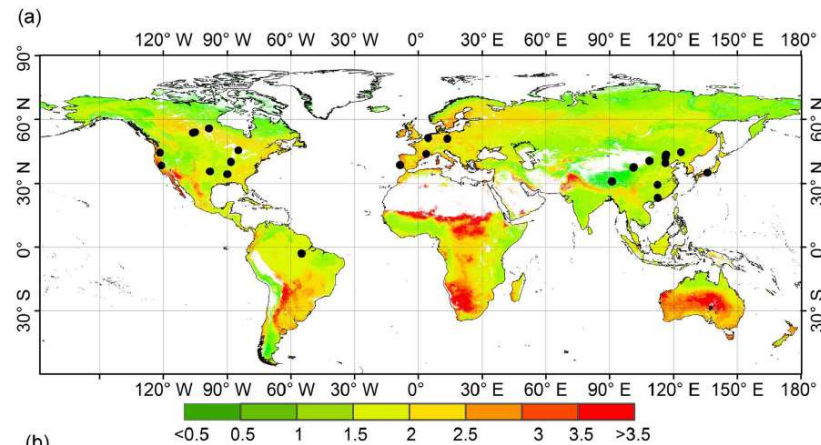
Sun et al., 2016, GEB



Tang et al., 2014, Scientific Reports

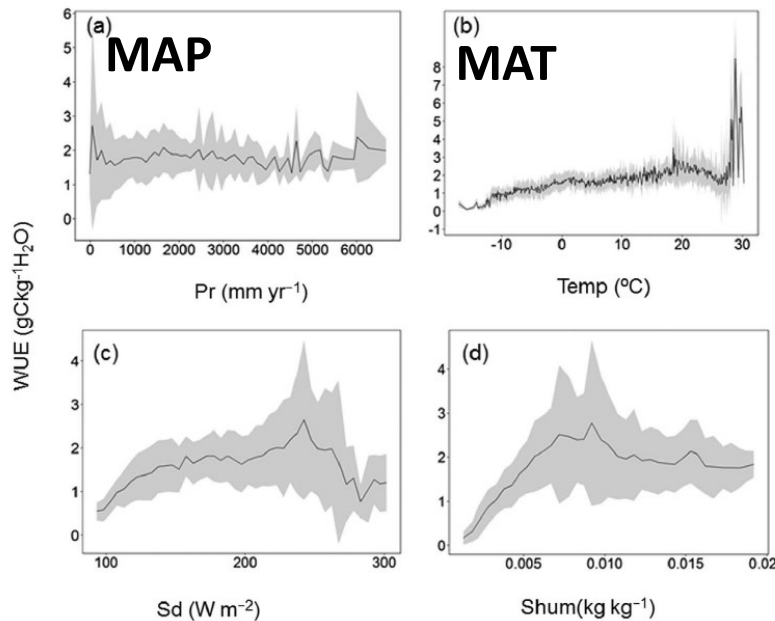
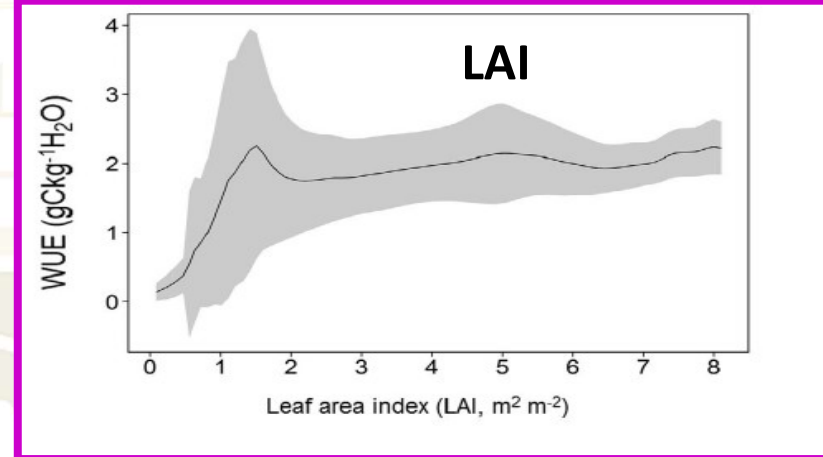
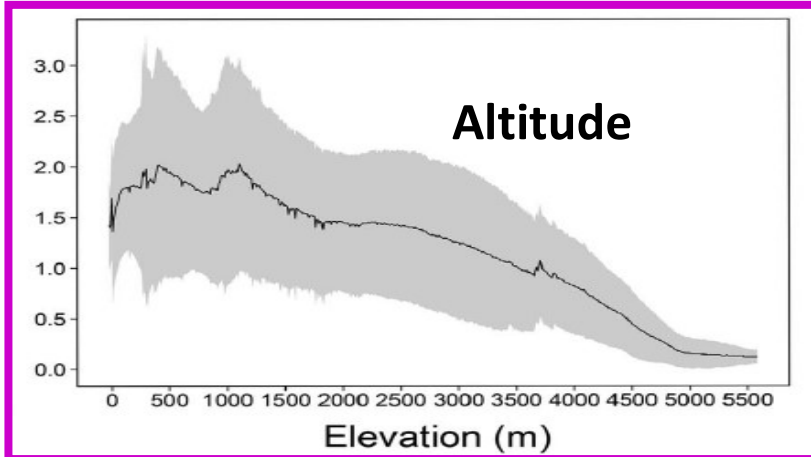
Data-driven or modeled GPP and ET

MODIS, JUNG's Data, process models



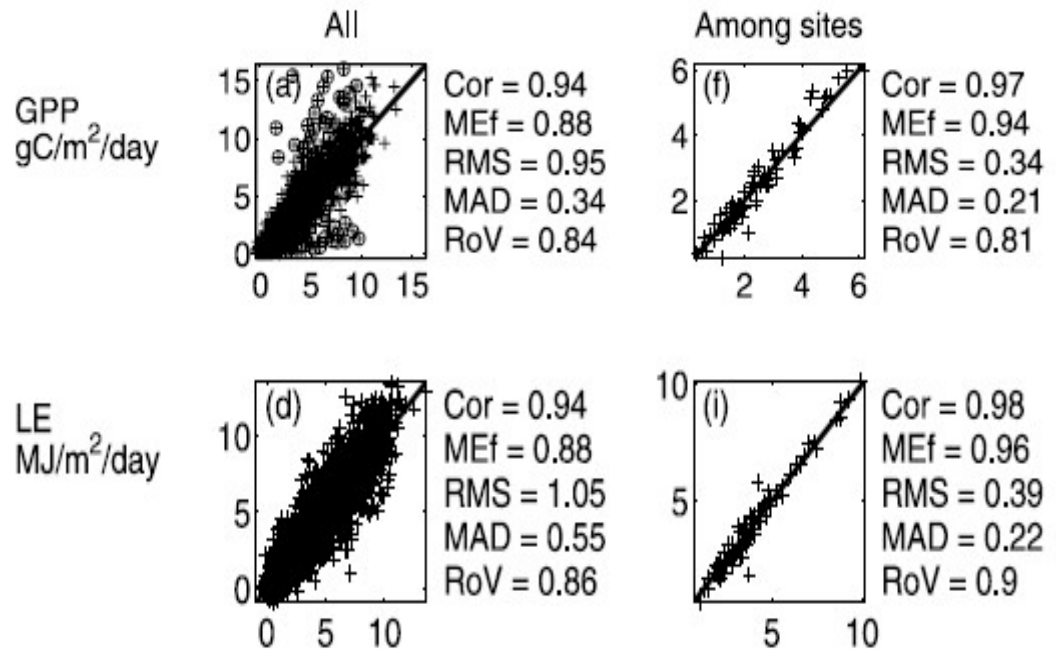
Xue et al., 2015, Ecosphere

Affecting factors



Many factors were found to affect the spatial variation of WUE





Jung et al., 2011, JGR

Uncertainties in data-driven or process-modelled GPP and ET



Uncertainties in WUE

Measuring GPP and ET



An alternative approach to unravel the spatial variation of WUE

FLUXNET

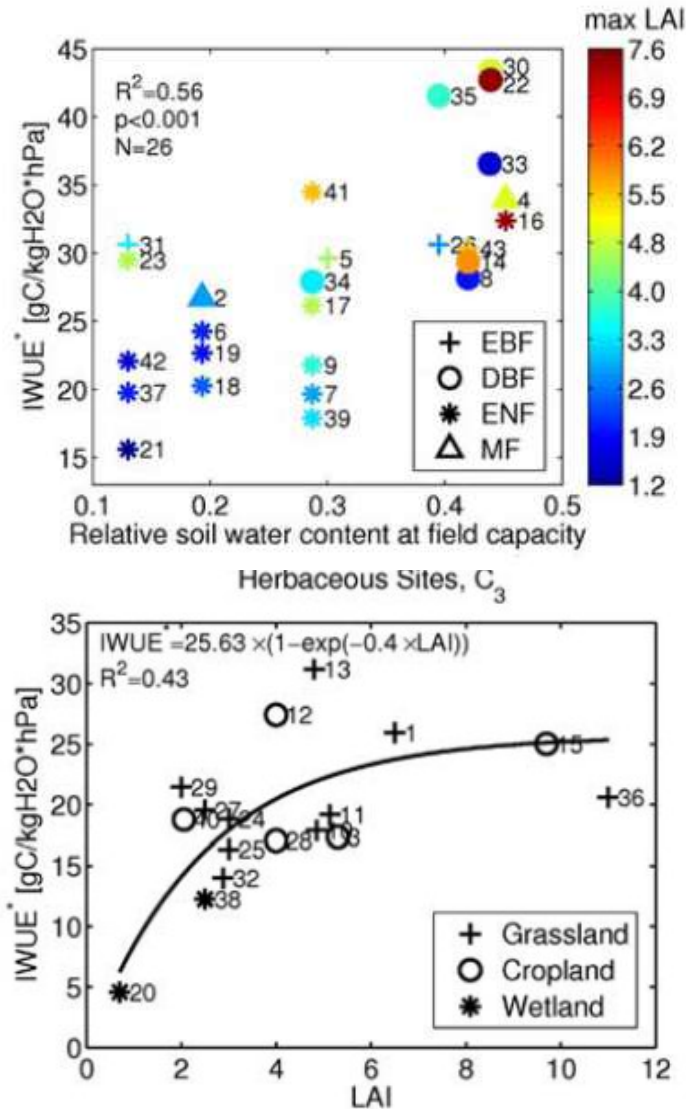


Table 2. Characteristics of Flux Tower Sites Used in This Study^a

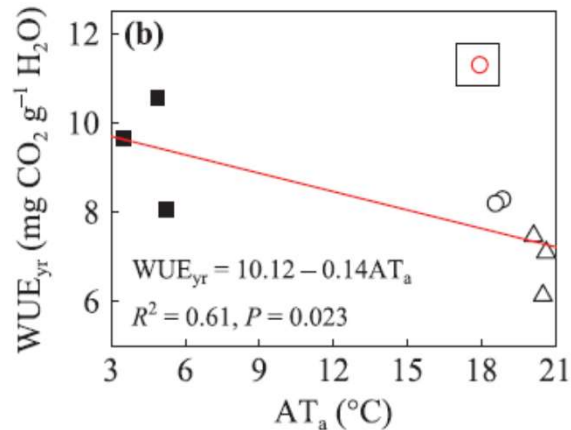
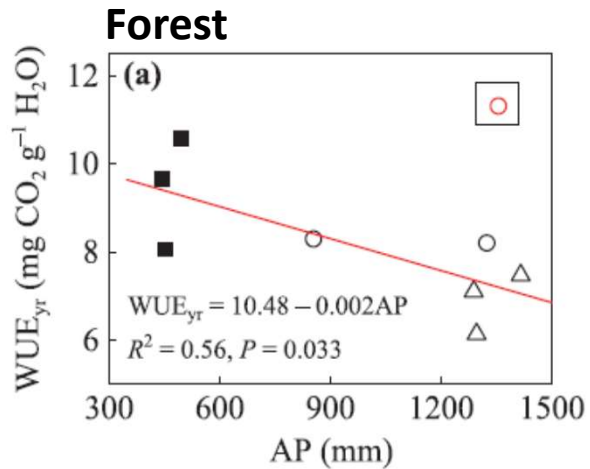
	Lat	Long	ID	Veg	LAI	Soil	WUE*	IWUE*	Reference
1	47.12	11.32	AT-Neu	GRA	6.5	0	3.79	25.94	Wohlfahrt et al. [2008a]
2	51.31	4.52	BE-Bra	MF	3	3	3.99	26.71	Carrara et al. [2004]
3	50.55	4.74	BE-Lon	CRO	5.3	0	2.83	17.35	Moureaux et al. [2006]
4	50.31	6	BE-Vie	MF	5.1	5	5.08	33.91	Aubinet et al. [2002]
5	-2.61	-60.21	BR-Ma2	EBF	4.7	6	2.82	29.6	n.a.
6	49.87	-125.29	CA-Ca2	ENF	2.2	3	3.06	24.26	Humphreys et al. [2005]
7	49.53	-124.90	CA-Ca3	ENF	3	1	3.53	19.65	Jassal et al. [2008]
8	53.63	-106.2	CA-Oas	DBF	2.1	10	3.41	28.18	Krishnan et al. [2006]
9	54	-105.12	CA-Obs	ENF	3.8	1	3.05	21.77	Krishnan et al. [2008]
10	47.29	7.7	CH-Oe1	GRA	4.85	0	2.86	17.88	Ammann et al. [2007]
11	31.52	122	CN-Do1	GRA	5.13	0	2.62	19.15	Wu et al. [2005]
12	51.10	10.91	DE-Geb	CRO	4	0	4.02	27.39	Anthoni et al. [2004b]
13	50.95	13.51	DE-Gri	GRA	4.8	0	4.35	31.17	Gilmanov et al. [2007]
14	51.08	10.45	DE-Hai	DBF	6	10	5.31	29.37	Knohl et al. [2003; Kutsch et al., 2008]
15	50.89	13.52	DE-Kli	CRO	9.7	0	3.58	25.01	n.a.
16	50.96	13.57	DE-Tha	ENF	7.6	5	4.55	32.4	Grünwald and Bernhofer [2007]
17	50.45	11.46	DE-Wet	ENF	4.75	1	5.42	26.17	Anthoni et al. [2004a]
18	39.35	-0.32	ES-ES1	ENF	2.63	3	2.77	20.27	Sanz et al. [2004]
19	61.85	24.3	FI-Hyy	ENF	2.1	3	3.61	22.68	Suni et al. [2003]
20	69.14	27.3	FI-Kaa	WET	0.7	0	1.23	4.58	Aurela et al. [2004]
21	67.36	26.64	FI-Sod	ENF	1.2	2	2.82	15.6	Thum et al. [2007]
22	48.67	7.07	FR-Hes	DBF	7.6	8	4.51	42.71	Granier et al. [2000]
23	44.71	-0.77	FR-LBr	ENF	4.8	2	2.63	29.47	Berbigier et al. [2001]
24	45.64	2.74	FR-Lq1	GRA	3	0	2.75	18.79	Allard et al. [2007]
25	45.64	2.74	FR-Lq2	GRA	3	0	2.42	16.36	Allard et al. [2007]
26	43.74	3.63	FR-Pue	EBF	2.9	7	3.14	30.61	Rambal et al. [2003]
27	46.69	19.60	HU-Bug	GRA	2.5	0	2.1	19.53	Nagy et al. [2007]
28	47.84	19.73	HU-Mat	CRO	4	0	2.32	17.05	Nagy et al. [2007]
29	41.90	13.61	IT-Amp	GRA	2	0	3.16	21.45	Wohlfahrt et al. [2008b]
30	41.85	13.59	IT-Col	DBF	5	8	6.07	43.39	Valentini et al. [1996]
31	41.70	12.38	IT-Cpz	EBF	3.5	2	3.51	30.61	Tirone et al. [2003]
32	46.01	11.05	IT-MBo	GRA	2.88	0	3	13.99	n.a.
33	44.69	11.09	IT-Non	DBF	1.7	11	3.15	36.56	n.a.
34	45.20	9.06	IT-PT1	DBF	3.5	1	2.98	27.91	n.a.
35	42.39	11.92	IT-Ro2	DBF	3.9	7	3.54	41.54	Tedeschi et al. [2006]
36	51.97	4.93	NL-Ca1	GRA	11	0	2.29	20.58	Jacobs et al. [2007]
37	52.17	5.74	NL-Loo	ENF	2.2	2	3.77	19.77	Dolman et al. [2002]
38	52.76	16.31	PL-Wet	WET	2.5	0	1.73	12.24	n.a.
39	64.11	19.46	SE-Fla	ENF	3.4	1	2.66	17.88	Lindroth et al. [2007]
40	36.61	-97.49	US-ARM	CRO	2.05	0	1.57	18.76	Fischer et al. [2007]
41	45.20	-68.74	US-Ho1	ENF	5.7	1	3.98	34.5	Hollinger et al. [2004]
42	29.75	-82.16	US-SP3	ENF	1.94	2	2.35	22.08	Clark et al. [2004]
43	-15.44	167.19	VU-Coc	EBF	5.65	10	3.17	30.33	Roupsard et al. [2006]

^aThe number in the first column is used in Figures 7 and 8 to indicate the sites. The station ID consists of two characters describing the country and 3 characters as abbreviation for the site name (cf. <http://www.fluxnet.ornl.gov/fluxnet/index.cfm>). Also shown are coordinates, vegetation class (EBF = evergreen broad-leaved forest, DBF = deciduous broad-leaved forest, ENF = evergreen needle-leaved forest, MF = mixed forest, GRA = grassland, CRO = cropland, WET = wetland), maximum leaf area index, soil texture type for forests according to [Cosby et al., 1984] (otherwise 0), mean WUE* [g C/kg H₂O], mean IWUE* [g C · hPa/kg H₂O], and a reference to site characteristics.

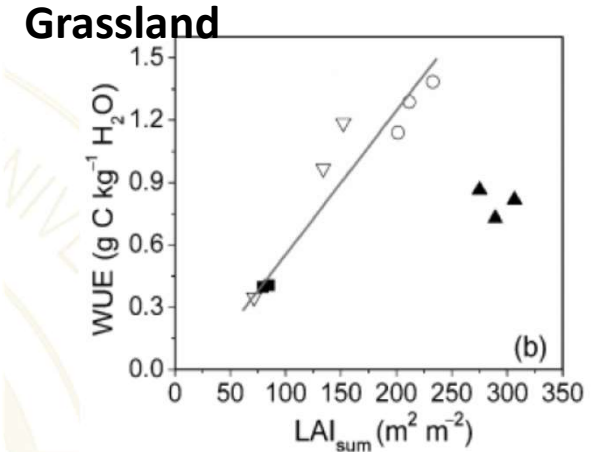
Covering few Chinese sites



China

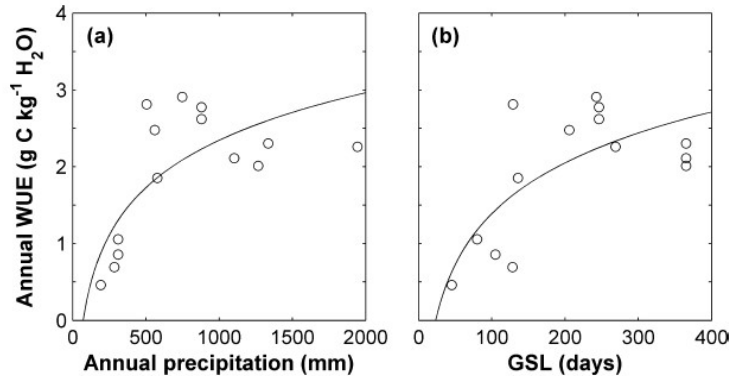


(Yu et al. 2008, New Phytologist)



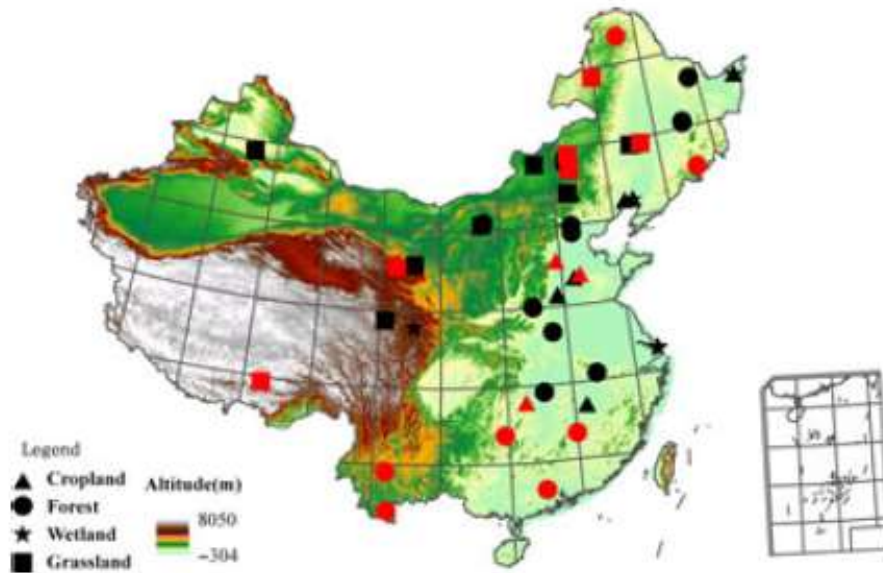
(Hu et al. 2008, GCB)

14 sites



(Xiao et al. 2013, AFM)

Inhibiting full understanding on the spatial variability of WUE in China based on measurements



(Yu et al. 2013, GCB)

Conducting eddy covariance measurements since 2002

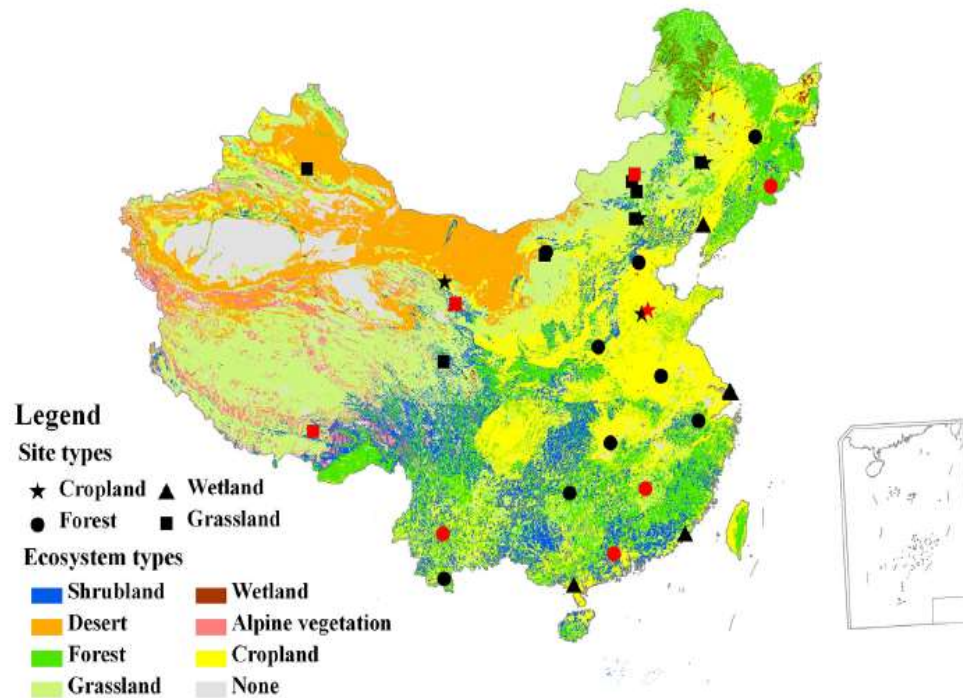
Accumulating a great deal of CO₂ and H₂O flux data

Objectives:

➤ How WUE spatially varied?

➤ Why WUE spatially varied?

How WUE spatially varied?



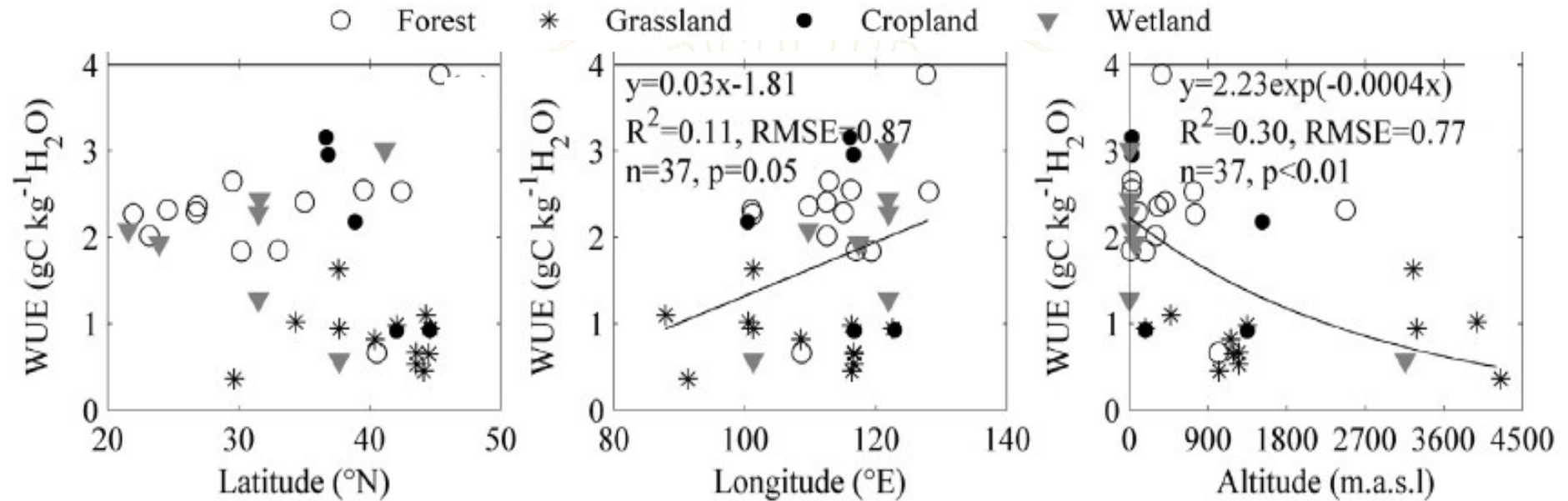
Selecting criteria

- Eddy covariance measurements
- Conducted for at least 1 year
- annual total GPP and ET were available
- GPP and ET were observed in the same year

Sites used in this study

Forests: 9; Grasslands: 9; Croplands:9; Wetlands: 6

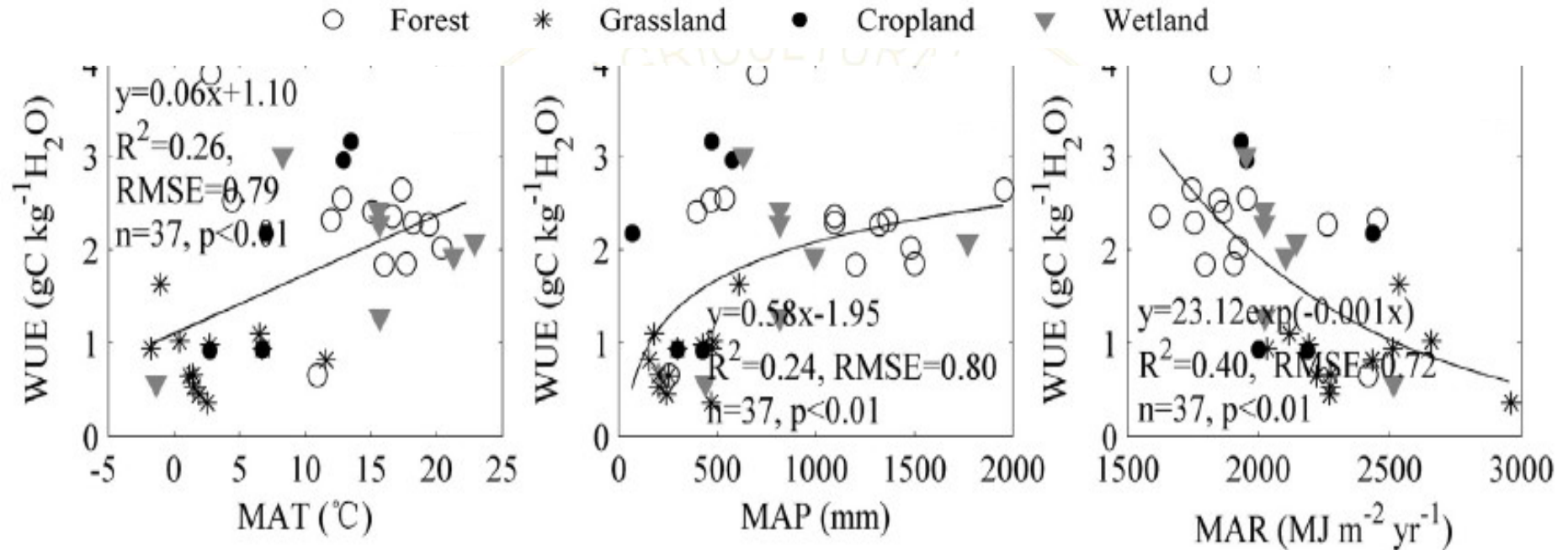
Geographical pattern



Geographical pattern of WUE in terrestrial ecosystems

- **Latitude: No significant**
- **Longitude: Slight increasing**
- **Altitude: Obvious decreasing**

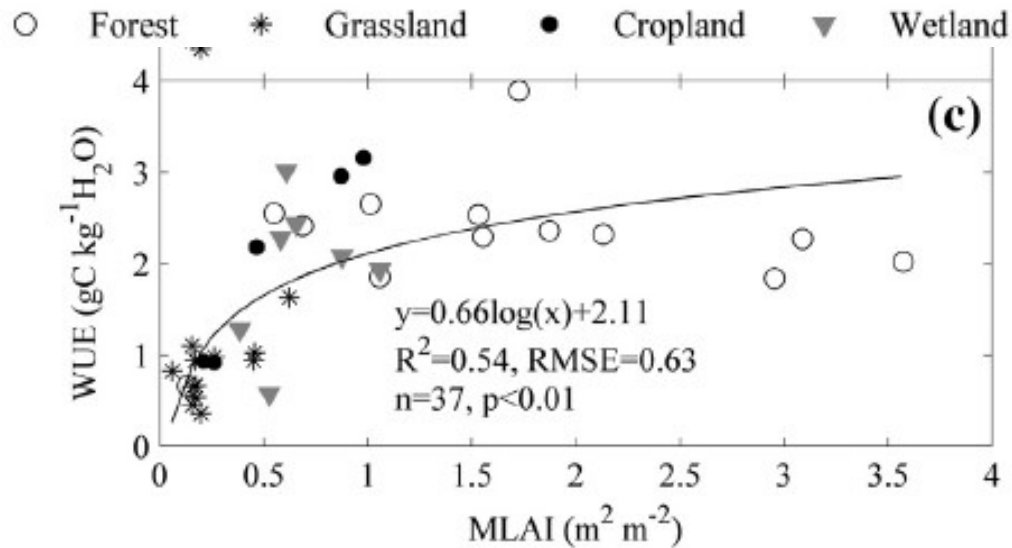
Climatic effects



Climatic effects on the spatial variation of WUE

- **MAT: Linearly increasing**
- **MAP: Logarithmically increasing**
- **PAR: Exponentially decreasing**

Biotic effect



➤ LAI: Logarithmically increasing

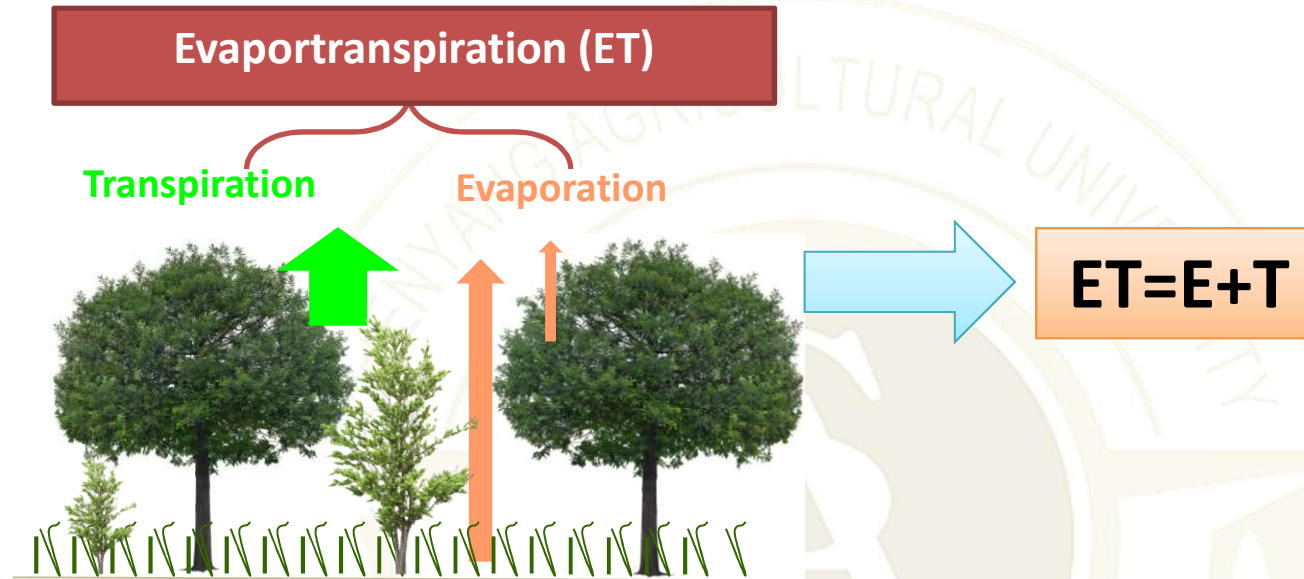
Biotic effect on the spatial variation of WUE

➤ Altitude and LAI explained most spatial variation of WUE

$$WUE = 0.58 \log(\text{MLAI}) - 0.0003\text{Altitude} + 2.32,$$

$$R^2 = 0.65, RMSE = 0.55$$

Why WUE spatially varied?



$$\text{WUE} = \frac{\text{GPP}}{\text{ET}} = \frac{\text{GPP}}{\text{T}} \times \frac{\text{T}}{\text{ET}}$$

The relationships between GPP/T and factors would be confounded by the effect of VPD on T [[Baldocchi et al., 1985](#); [Beer et al., 2010](#)]

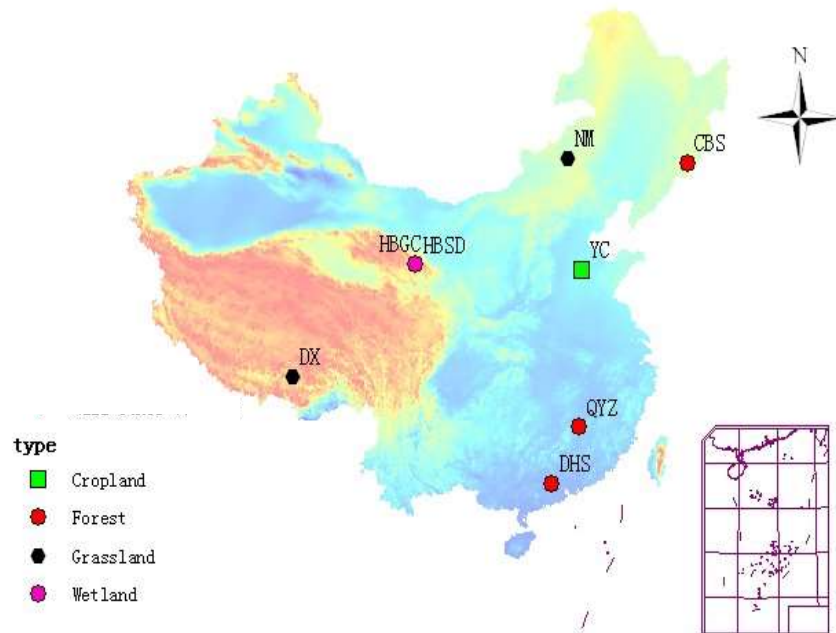
The effect of VPD on T would only also occur in the **growing-season**

$$IWUE = GPP \times VPD_{gs} / T$$

$$WUE = IWUE \times 1 / VPD_{gs} \times T / ET$$

Analysing factors affecting the spatial variations of WUE components would reveal why various factors affected the spatial variation of WUE

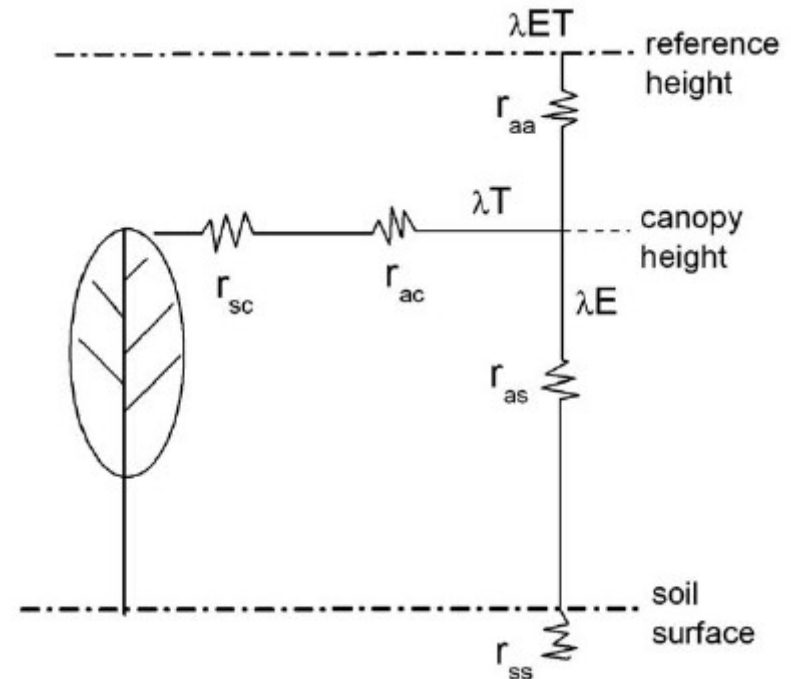
Materials



ChinaFLUX observation sites

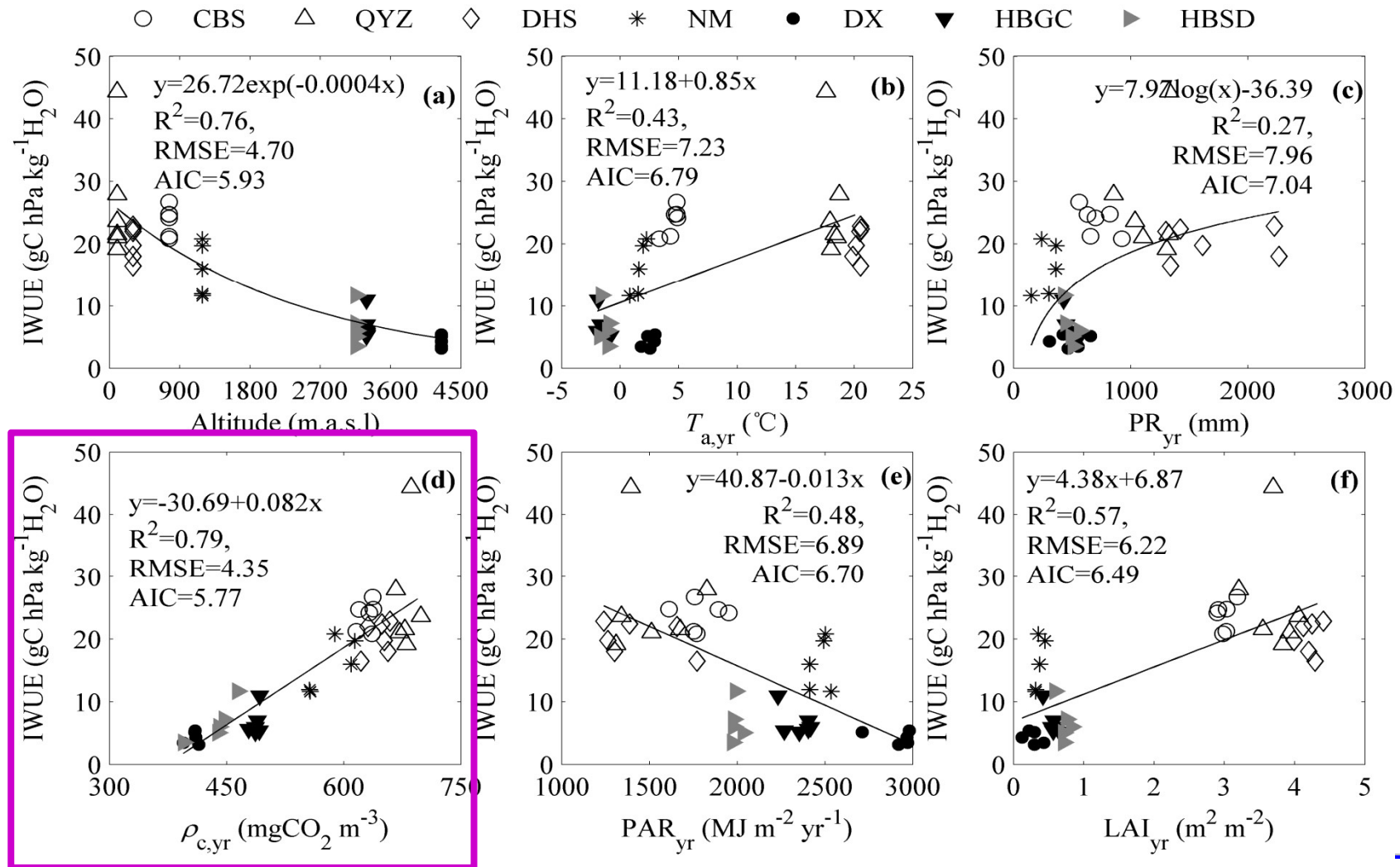
Using S-W model
separating ET into E and T

Eddy covariance measured GPP and ET



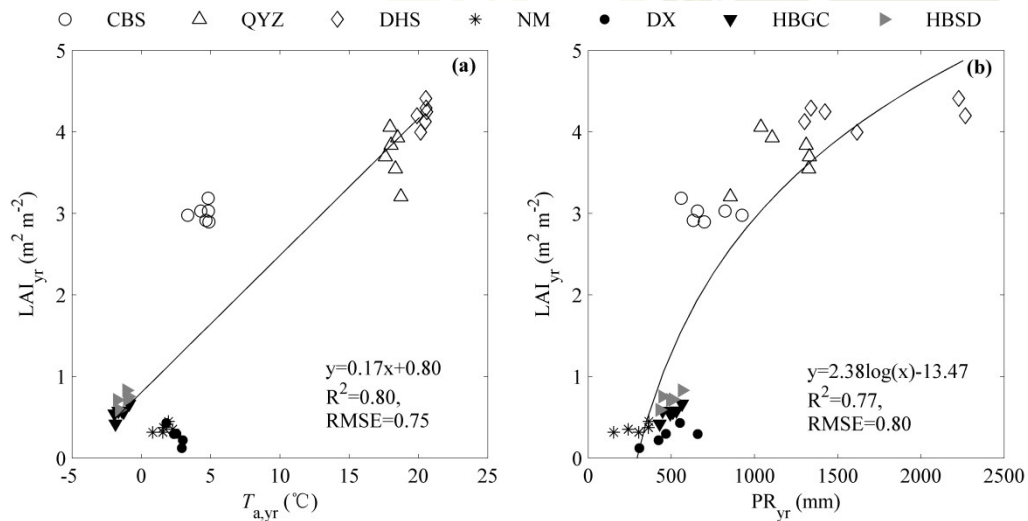
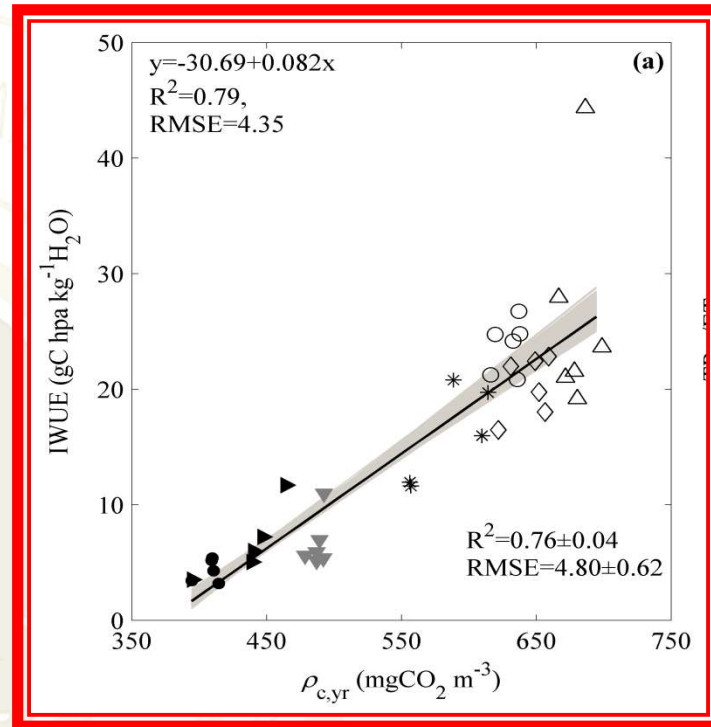
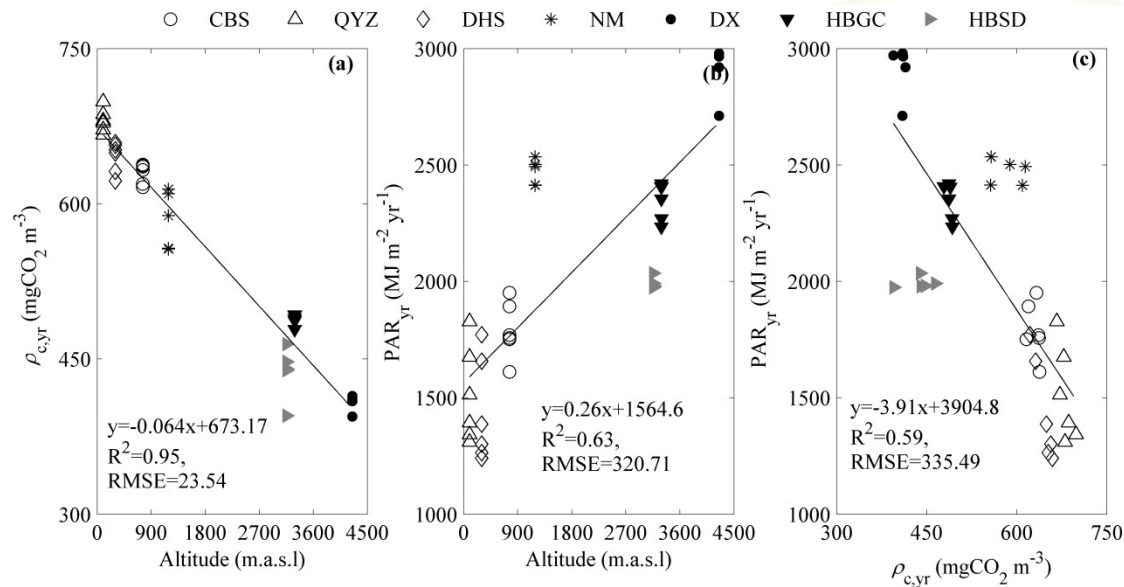
Scheme of Shuttleworth-Wallace model

IWUE



Factors affecting the spatial variability of IWUE

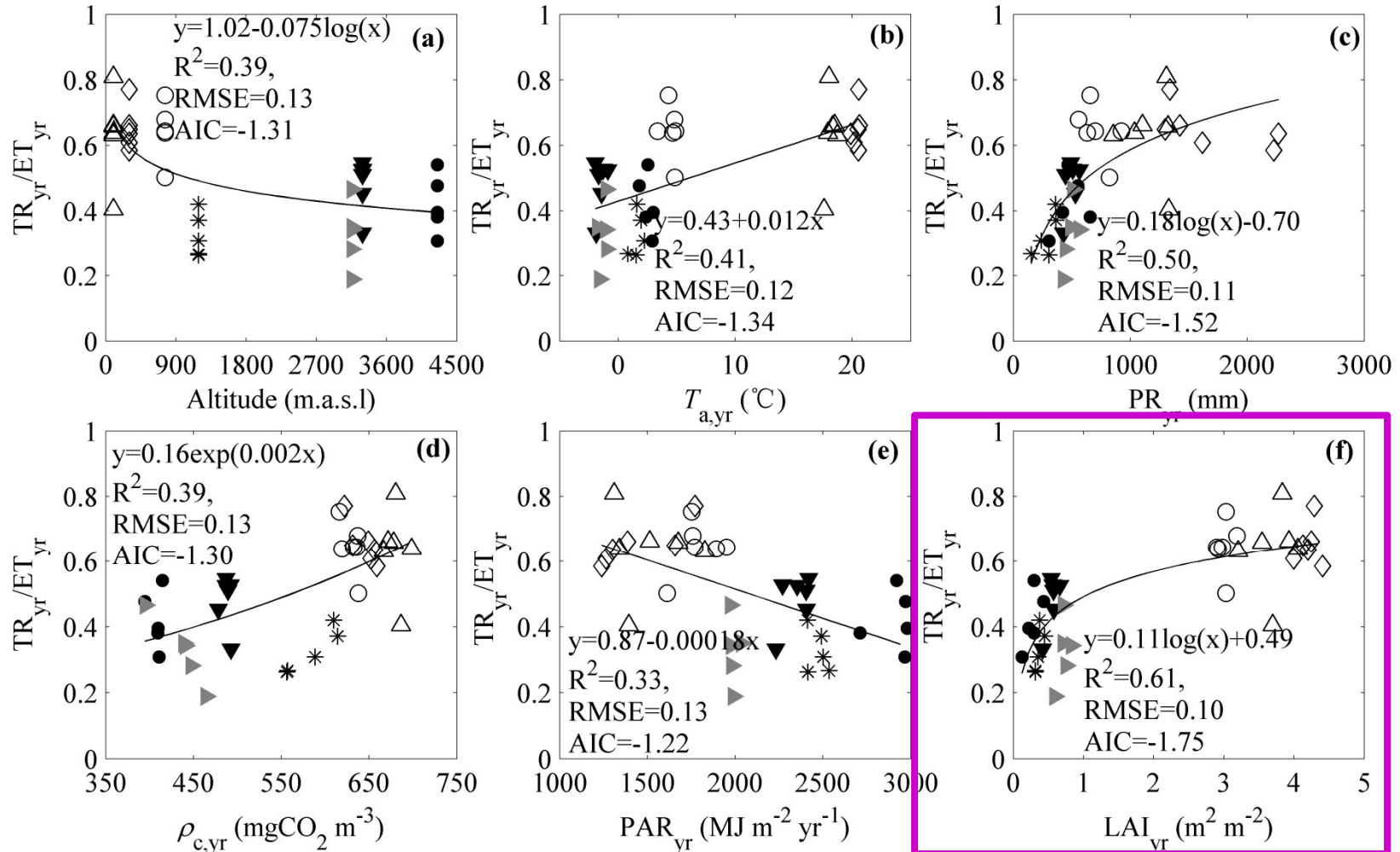




$\rho_{c,yr}$ dominated the spatial variation of IWUE

T/ET

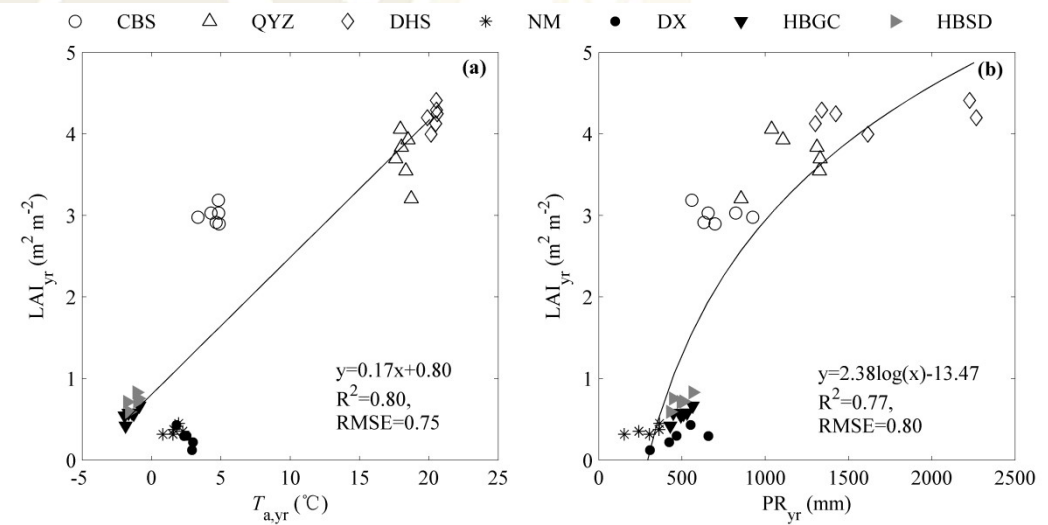
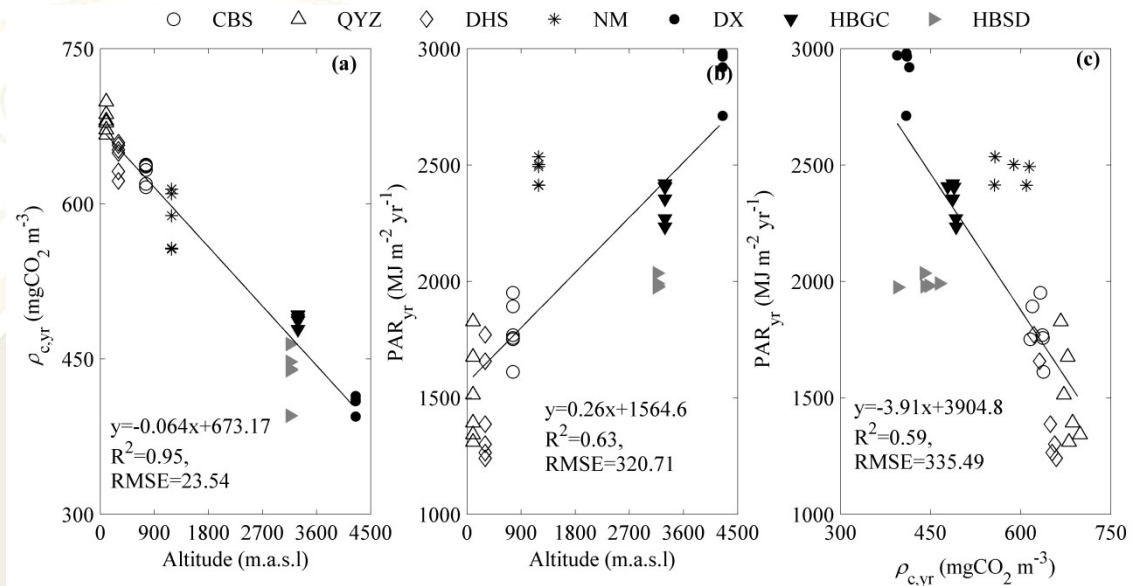
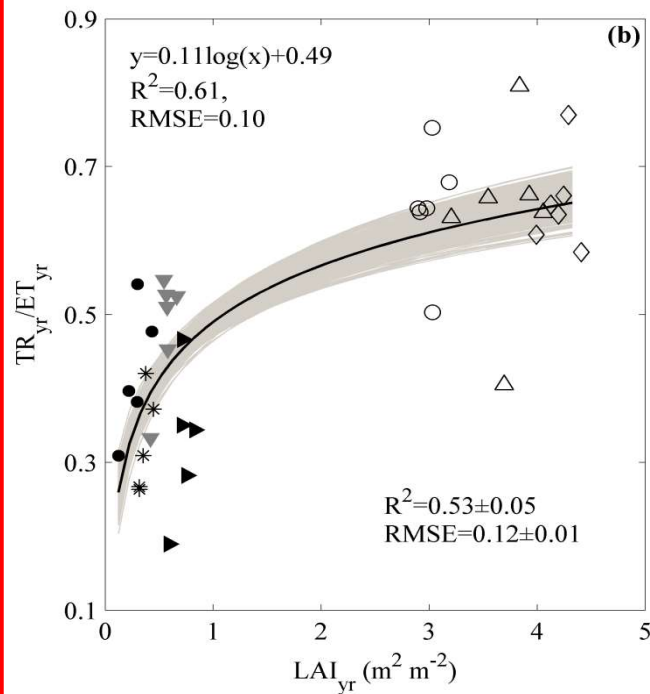
○ CBS △ QYZ ◇ DHS * NM ● DX ▼ HBGC ► HBSD



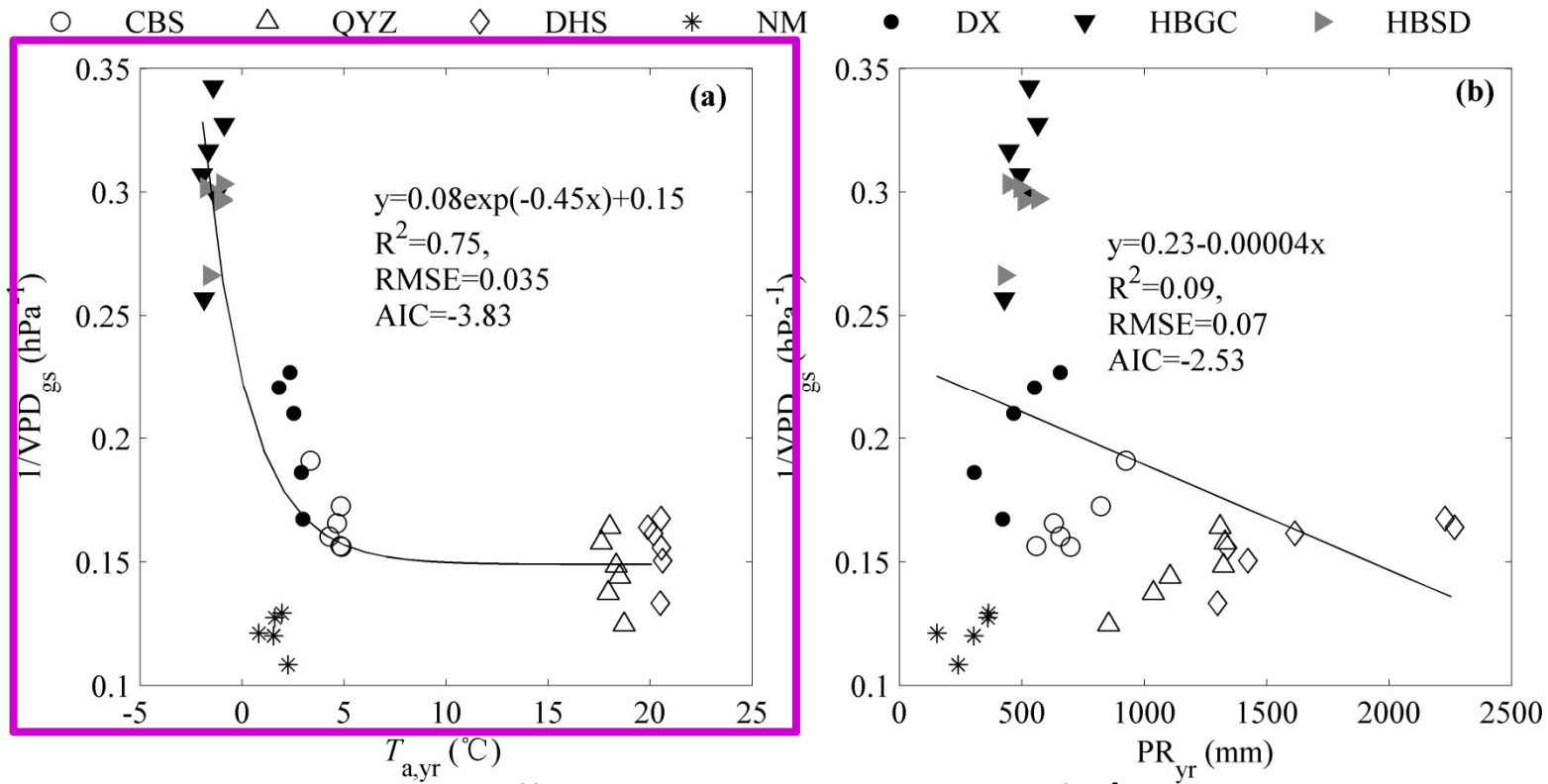
Factors affecting the spatial variability of T/ET



LAI closely related to the spatial variation of T/ET



1/VPD_{gs}



Factors affecting the spatial variability of 1/VPD_{gs}

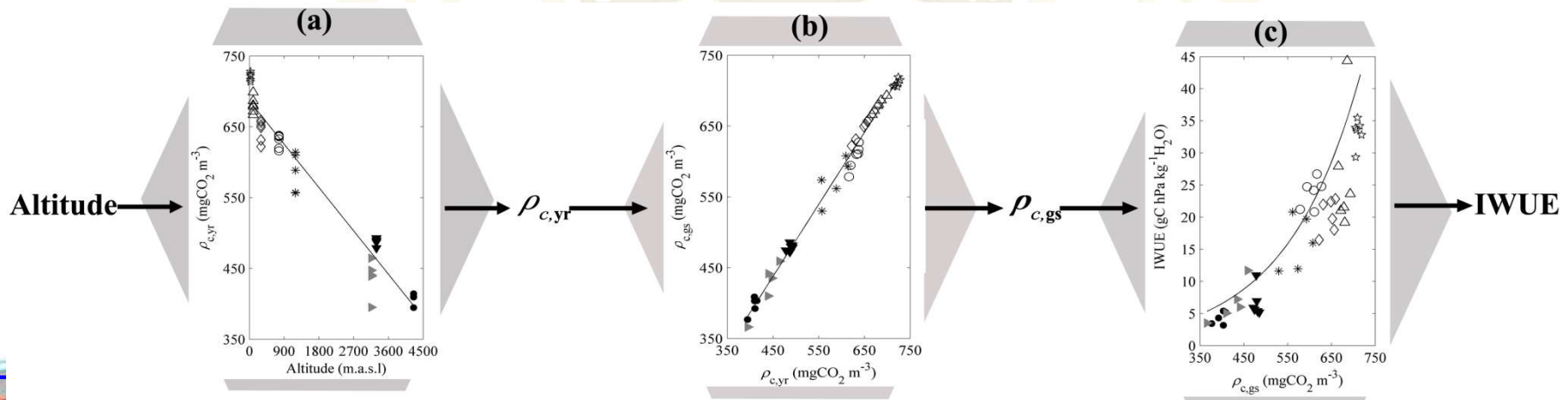
MAT dominated the spatial variation of 1/VPD_{gs}

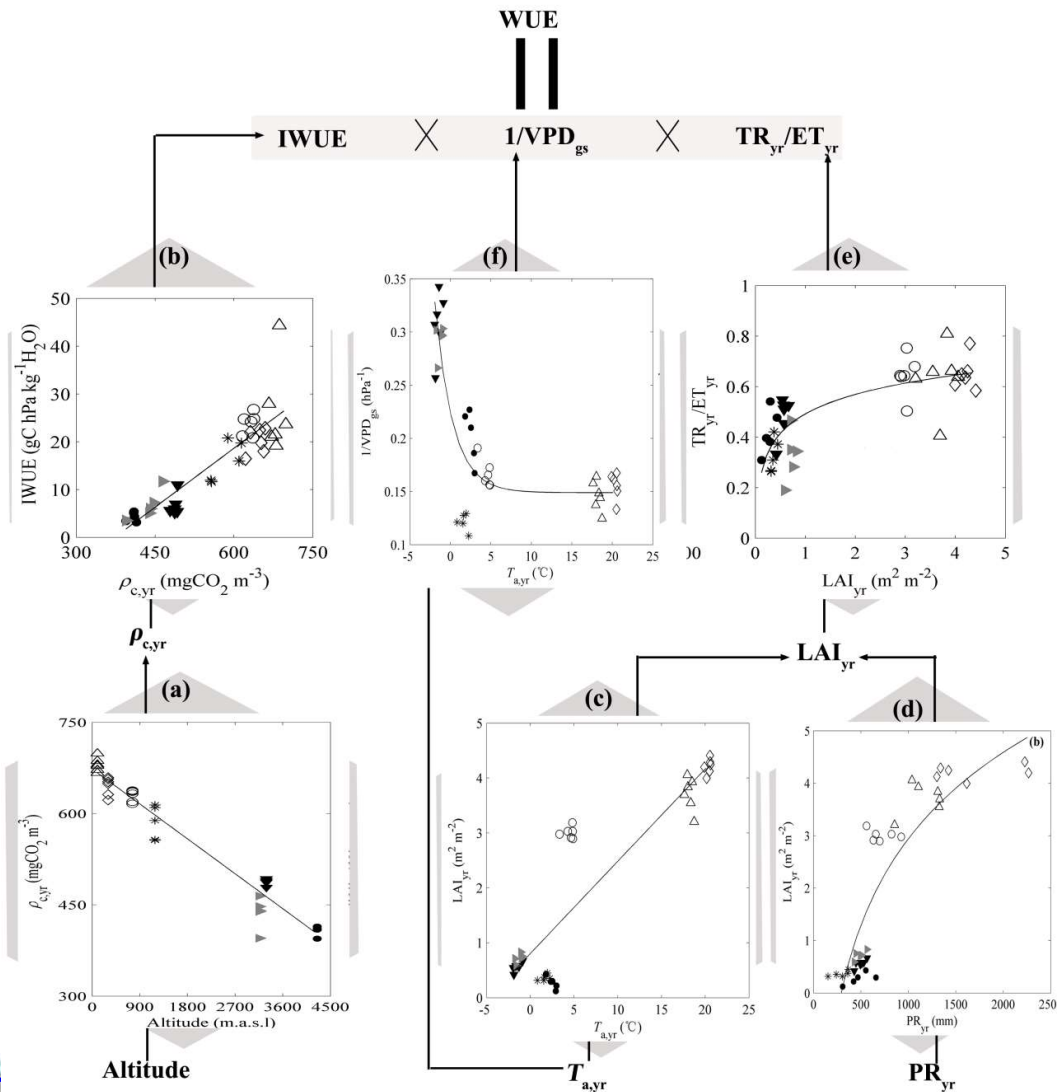
Mechanisms

$$GPP = G_c \times D\rho_c = G_c \times (\rho_c - \rho_{ci})$$

$$Tr = G_w \times D\rho_c = G_w \times (\rho_{wi} - \rho_{wa})$$

$$IWUE \propto \frac{\rho_{c,gs} (1 - \rho_{ci} / \rho_c)_{gs}}{R}$$





- **Altitude** affected the spatial variation of WUE through the effect of $\rho_{c,yr}$ on **IWUE**;
- **MAP** influenced the spatial variation of WUE through the effect of **LAI** on **T/ET**
- **MAT** affected the spatial variation of WUE through its effect on **T/ET** and **1/VPD_{gs}**.
- WUE was the product of **IWUE**, **T/ET**, and **1/VPD_{gs}**.



Summary



- WUE exhibited a decreasing trend with the increasing altitude, but increased with MAT, MAP, and LAI.
- The spatial variation of WUE could be depicted by the equation of altitude and mean annual leaf area index.
- $\rho_{c,yr}$ dominated the spatial variation of IWUE through the altering altitude while LAI affected the spatial variation of T/ET, but MAT determined the spatial variation of $1/VPD_{gs}$.
- WUE was the product of IWUE, T/ET, and $1/VPD_{gs}$.





Acknowledgements

- I am very grateful for data provider in ChinaFLUX providing eddy covariance observation data, including sites as CBS, QYZ, DHS, NM, DX, HBGC, HBSD, YC.
- Many thanks to the finance support by the National Key Research and Development Program and National Natural Science Foundation of China

Thank you for your attention!



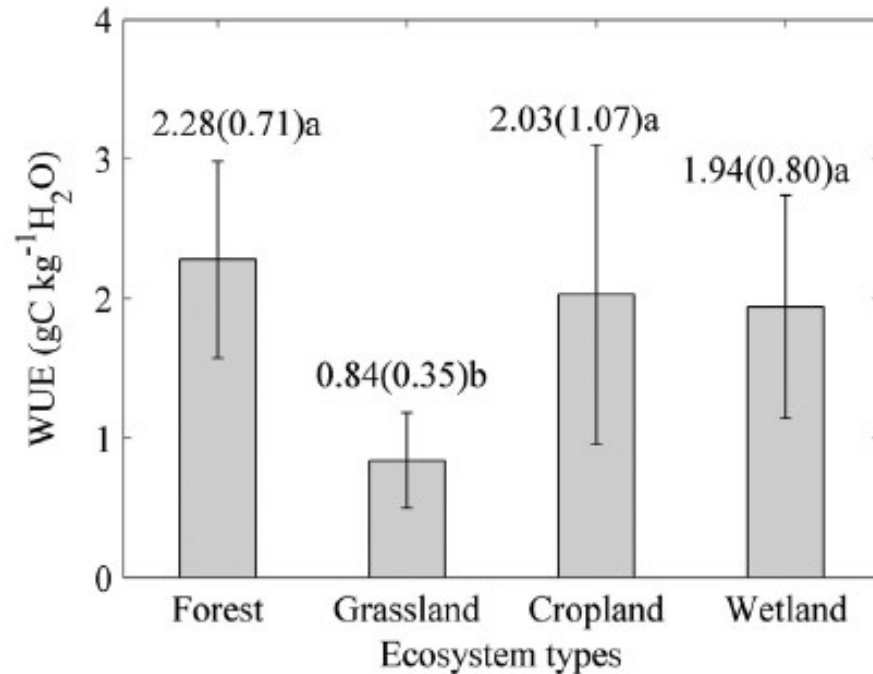
Further implications



- CO₂ not only affects the temporal variation of WUE, but also affects its spatial variation.
- $\rho_{c,yr}$ may be a potential climatic factor affecting the spatial variation of carbon fluxes



Statistical values



- **Forest, Cropland, Wetland** were comparable
- **Grassland** was significantly lower than other ecosystems

Statistical values of WUE among ecosystem types