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Water scaling of ecosystem carbon cycle feedback to climate warming

Author(s):
Quan, Quan; Tian, Dashuan; Luo, Yiqi; Zhang, Fangyue; Crowther, Tom W.; Zhu, Kai; Chen, Han Y. H.; Zhou, Qingping; Niu, Shuli

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It has been well established by field experiments that warming stimulates either net ecosystem carbon uptake or release, leading to negative or positive carbon cycle–climate change feedback, respectively. This variation in carbon-climate feedback has been partially attributed to water availability. However, it remains unclear under what conditions water availability enhances or weakens carbon-climate feedback or even changes its direction. Combining a field experiment with a global synthesis, we show that warming stimulates net carbon uptake (negative feedback) under wet conditions, but depresses it (positive feedback) under very dry conditions. This switch in carbon-climate feedback direction arises mainly from scaling effects of warming-induced decreases in soil water content on net ecosystem productivity. This water scaling of warming effects offers generalizable mechanisms not only to help explain varying magnitudes and directions of observed carbon-climate feedback but also to improve model prediction of ecosystem carbon dynamics in response to climate change.

RESULTS AND DISCUSSION
To test the above hypothesis and to understand the underlying mechanisms, we conducted a field manipulation experiment in an alpine meadow of the Qinghai-Tibetan Plateau. The unique feature of this alpine meadow is a wide span of SWC from extremely wet to dry conditions and the high warming sensitivity (15, 16). Our experiment consisted of three temperature treatments, including ambient (C), +1.5°C (W1.5), and +2.5°C (W2.5) soil temperatures (STs) at a depth of 10 cm. On average, across 3 years, warming significantly decreased SWC by 2.3 and 5.8%, respectively, with W1.5 and W2.5 treatments (fig. S1). Measured NEP, gross ecosystem productivity (GEP), and ecosystem respiration (ER) all showed diverse responses to warming across years (fig. S2) and among the seasons within the year (fig. S3). Across all the measurements, GEP, ER, and NEP followed hump-shaped response surfaces with ST and SWC (Fig. 2). The response surface resulted from a quadratic relationship of GEP, ER, and NEP with SWC and an exponential relationship with ST (Eq. 4 in the “Statistical analysis” section and fig. S4). A clear ridge along an SWC of 29.9 ± 2.4% (i.e., the optimum SWC) emerged from the hump-shaped response surfaces. GEP, ER, and NEP increased with SWC below its optimum but decreased above the optimum. The SWC optimum tended to shift toward larger values as temperature increased (Fig. 2).

The hump-shaped response surfaces can help us to explain diverse warming effects on GEP, ER, and NEP with changing SWC (Fig. 3). When SWC was higher than its optimum (the ridge of the hump-shaped surface), warming-induced increase in ST (red arrows) and decrease in SWC (black arrows) both promoted GEP (blue dashed arrows represent their combined effects). When SWC was below the optimum, the combined positive impact of increasing temperature with negative impact of decreasing SWC under experiment warming resulted in either an increase or a decrease in GEP. As GEP response
was always greater than that of ER in this study, the response pattern was similar for both GEP and NEP (Fig. 3). Warming alone consistently promotes plant growth and enhances GEP under adequate moisture availability (10, 17). Warming also accelerates microbial decomposition of soil organic C and thus increases ER (18–20). Under water deficits, the negative impacts of warming-induced droughts limit both carbon uptake and ER (5, 10, 21, 22). Since plant photosynthesis is more sensitive to drought than respiration (23, 24), GEP decreased more than ER, leading to a decrease in NEP and a positive carbon-climate feedback under low SWC.

We used a nonlinear statistical model to discern the relative effects of warming-induced changes in temperature and SWC on GEP, ER, and NEP. Our analysis demonstrated that increasing ST under warming treatments alone always raised GEP, ER, and NEP across the full range of SWC (Fig. 4). However, warming-induced decreases in SWC stimulated GEP, ER, and NEP in its high range but depressed them in the low range (Fig. 4). Changes in these carbon fluxes formed linear scaling relationships with SWC. As a consequence, the warming effect was linearly scaled up or down with SWC, depending on which ranges the warming-induced decreases in SWC fell. Below the optimum, warming-induced decline in SWC reduced carbon fluxes to offset the stimulation caused by increasing ST, leading to either decreases in GEP, ER, and NEP under severe dry conditions or minor increases in these carbon fluxes under moderate dry conditions (Fig. 4). Above the optimum, warming-induced changes in SWC and ST both stimulated GEP, ER, and NEP. Their combination amplified each other to produce a much stronger stimulation than their individual effects (Fig. 4). While the patterns of water scaling were similar between the two levels of temperature treatments, the magnitudes of impacts were higher under the W2.5 treatment than those under the W1.5 treatment (Fig. 4).

To test whether this water scaling pattern is generalizable across ecosystems, we conducted a global meta-analysis. We examined warming impacts on NEP in relation with ambient precipitation, due to fewer data points reported on warming-induced changes in SWC in the literatures from the study sites (table S1). However, precipitation is a good proxy for SWC at both site levels (figs. S5 and S6) and the global scale as shown in another synthesis study (25). In general, warming-induced changes in NEP were highly correlated with ambient precipitation (Fig. 5, P = 0.003) but not with the magnitude or duration of experimental warming (table S2) or mean annual temperature (table S2). This relationship holds even when experimental data only from the temperate zone were used and those from other climate zones (e.g., tundra ecosystems) were excluded. Overall, warming significantly reduced NEP under low precipitation but increased it under high precipitation.

This water scaling can well explain conflicting results reported from previous case studies on warming-induced changes in ecosystem carbon fluxes (26, 27). Increases in plant growth, GEP, ER, or NEP under experimental warming were mostly detected in wet ecosystems, such as subtropical forests with a mean annual precipitation (MAP) of 1778 mm (28), temperate spruce forests with a MAP of 1480 mm (5), and alpine meadows with a MAP of 966 mm (29). In contrast, the warming-induced decreases in these ecosystem carbon fluxes were mainly reported in dry ecosystems, such as semiarid grasslands with a MAP of 375 and 241 mm (24, 30), high arctic tundra with a MAP of 197 mm (19), and high arctic polar semidesert with a MAP of 122 mm (10).

Fig. 1. Conceptual diagram of warming effects on ecosystem C fluxes above and below the SWC optimum. The red arrows represent the directly positive warming effect on ecosystem C fluxes. The black arrows represent the effect of warming-induced changes in SWC on C fluxes, which is the indirect effect of warming. Below the SWC optimum, the warming-induced decrease in SWC reduces C fluxes; thus, the black arrow points downward along the SWC-C flux response curve. Above the SWC threshold, warming-induced water loss increases C fluxes; thus, the indirect warming effect enhances C fluxes, and the black arrow points upward along the SWC-C flux response curve. The blue dashed arrows represent the final change direction of C fluxes under the combination of both direct and indirect effects of warming. The photograph depicts our field experimental plots at the study site. (Photo credit: Q.Q., Institute of Geographic Sciences and Natural Resources Research, Chinese Academy of Sciences).

Fig. 2. Response surfaces showing the relationships between ST and SWC versus GEP, ER, or NEP across plots and years. Observed values (black crosses) are the means of five replications of C fluxes, and modeled values (colored surfaces) are predictions from the models fitted with observations.
This study, to our knowledge, reconciles mechanisms revealed from site-level experiments with patterns of water scaling of warming effects at global scales. Our experiment in the Qinghai-Tibetan Plateau revealed a flipping pattern, as did from the boreal forest experiment (12), that warming effect was negative under dry conditions but positive under wet conditions. This flipping pattern is confirmed by our meta-analysis to be generalizable across experiments in different ecosystems. Our combined study suggests that water availability represents a strong, consistent mechanism underlying diverse, sometimes contradictory, responses of carbon processes to climate warming observed from different ecosystems.

Beyond revealing a generalizable pattern of water scaling, our study has also developed a continuous (i.e., quadratic) function (Eq. 4 in the "Statistical analysis" section) to represent scaling of carbon fluxes with SWC. This function allowed us to identify thresholds of SWC between stimulating and inhibiting warming effects on carbon fluxes (Figs. 2 and 3). The threshold becomes higher under high (W2.5) than low (W1.5) temperature treatments. This continuous function also allowed us to attribute observed ecosystem responses to direct warming effects versus indirect warming effects through changes in SWC (Fig. 4). Direct warming effects at our alpine meadow site were always positive along SWC gradients for all the GEP, NEP, and ER, whereas indirect warming effects through changes in SWC were positive under wet conditions but negative under dry conditions. This water scaling function has the potential to improve model prediction of carbon cycle feedback to climate change. Models have used a variety of response functions, such as linear (31), Gaussian (32), exponential (33), and reverse exponential (34), to represent carbon cycle responses to SWC. These functions are apparently not sufficient to represent a full spectrum of C cycle responses to water availability. For example, the linear response function is likely to work well when one ecosystem is prevailingly under either dry or wet condition. In comparison, this study reveals that a continuous function is needed to represent a full spectrum of hydrological regulation of the C cycle and its feedback to climate warming to improve model prediction of C cycle responses to climate warming (35).

Our findings shed light on future climate-carbon feedback at least in a couple of ways. Future warmer conditions will lead to variable feedbacks of ecosystems to climate warming, depending on moisture conditions. Ecosystems in very dry regions most likely decrease carbon uptake under warming conditions and thus cause a positive feedback to climate warming. In contrast, ecosystems in wet regions possibly generate a negative feedback. Moreover, warmer climate will result in chronically lowering soil moisture. Lowering soil moisture will stimulate warming effects in wet regions but exert a strong braking effect on, or even reverse, the potential benefit of climate warming on ecosystem carbon uptake in dry regions. Thus, this and other studies suggest precipitation thresholds to regulate NEP variation at regional and continental scales (36).

In summary, our and many other field warming experiments provide compelling evidence on water regulation of ecosystem feedback to climate warming at both temporal and spatial scales. The revealed water scaling represents a generalizable pattern for understanding ecosystem-climate feedback from local to global scales. A continuous function of water scaling is needed to describe observed shifts of carbon-climate feedback from being negative to positive along the full spectrum of water availability at our experimental site and likely other sites. This water scaling pattern is generally supported by the global meta-analysis that warming stimulates net ecosystem carbon release (i.e., positive carbon cycle feedbacks to climate warming) in low-precipitation conditions.
regions but enhances carbon uptake (i.e., negative feedbacks) in high-precipitation regions. This water scaling pattern enriches mechanistic understanding of temperature-moisture interactions in affecting ecosystem carbon cycling and its feedback to climate change and has the potential to improve model prediction.

MATERIALS AND METHODS
Study area for the warming experiment
The experimental study was conducted in an alpine meadow of the Eastern Qinghai-Tibetan Plateau (32°48′N, 102°58′E), which locates in Hongyuan County, Sichuan of China, at an altitude of approximately 3500 m. Over the past 60 years, the MAP is 753 mm, with approximately 80% occurring during May to September, and the mean annual temperature is 1.1°C, with January as the coldest month (−10.3°C) and July as the hottest month (10.9°C). The soil at the study site is classified as Cryumbrept following the U.S. Soil Taxonomy (37, 38). Plant species in this alpine meadow are dominated by *Deschampsia caespitosa* (Linn.) Beauv., *Koeleria cristata* (Linn.) Pers., *Gentiana sino-ornata* Balf. f., *Potentilla anserina* L., and *Anemone rivularis* Buch.-Ham.

Experimental design
We used random block design with three warming treatments and five replications for each treatment in this study. Three 3 m by 2 m plots were laid out in each of the five blocks and randomly assigned to the three treatments of control (C), low-level warming (W1.5), and high-level warming (W2.5). The warmed plots were continuously heated, proceeding in June 2014, via 165 cm by 15 cm infrared radiators (MSR-2420, Kalgo Electronics Inc., Bethlehem, Pennsylvania, USA), which were suspended in the center of the plot, at 1.5 m above ground level. The heaters for the W1.5 treatments were set at an output power of approximately 1000 W, with the expectation of a 1.5°C increase in ST at 10-cm depth, while the heaters for the W2.5 treatments were set at an output power of approximately 2000 W, with the expectation of a 3°C increase in ST. In each control plot, we suspended a dummy heater whose appearance is identical to the infrared radiator at the same height to simulate the shading effect. The adjacent plots were 3 m apart.

Measurement of ecosystem CO₂ fluxes
We measured NEP and ER twice per month during the growing season on clear days from June to September in 2014 to 2016. In May 2014, we installed a 0.5 m by 0.5 m square aluminum frame into the soil at the depth of 3 cm in each plot, with a distance of at least 30 cm from the perimeter of the plot, to seal the canopy chamber (0.5 m by 0.5 m, polymethyl methacrylate) to the soil surface and provided a plane interface between them. Care was taken to minimize soil disturbance during the installation. NEP and ER were measured using an infrared gas analyzer (LI-6400XT, LI-COR Environmental, Lincoln, Nebraska, USA), which was attached to the transparent canopy chamber. During measurements, two small fans were installed diagonally inside the chamber and fanned continuously to mix the atmosphere. Consecutive recordings of CO₂ concentrations were obtained once every 10 s in 80 s. NEP was calculated by the slope between recording time and concentrations [see the detailed method in (11)]. Right after the NEP measurements, the chamber was lifted up to exchange air with the outside. Then, we covered an opaque cloth on the chamber and repeated the measurement to obtain ER. GEP was calculated as the difference between NEP and ER.

ST (°C) was measured at a 10-cm depth using a thermocouple probe concurrently with the ecosystem CO₂ fluxes measurements. The 10-cm SWC (% V) was measured using a time domain reflectometry instrumentation (TDR 100, Spectrum Technologies Inc., Chicago, USA). All measurements were performed at the same time with C fluxes measurement.

Meta-analysis
To analyze the effect of precipitation on warming-induced changes in NEP at global scale, we synthesized various warming experiments involved in terrestrial ecosystems. First, we used “Web of Science” and “Google Scholar” to search peer-reviewed literatures that investigated NEP response under experimental warming during 1900–2019. Then, we screened the papers for analysis based on the criteria as follows. (i) Field studies must include a control and warming treatment under the same condition between them. (ii) The variable of NEP is shown by its mean and sample size. (iii) Experimental method needs to be explicitly described as well, such as warming magnitude, experimental duration, and warming method. Data shown in figures were extracted with the Engauge Digitizer (Free Software Foundation Inc., Boston, MA, USA). If multiple years’ data were reported in the same experiment, we only selected the latest measurement to assure statistical independence between observations. Overall, we established a global dataset composed of 34 independent experiments (including this study), with warming magnitude ranging from 0.2° to 3°C and experimental duration from 1 to 9 years. Since this study examines the response of NEE, we only included the direct measurement of NEE that was conducted by chambers in herbaceous ecosystem, including tall-grass prairie, temperate steppe, peatland, fens, alpine meadow, and tundra, with precipitation range from 100 to 914 mm and temperature from −14.6° to 16.3°C (table S1).

Warming effect was calculated as (NEP_treatment − NEP_control) / NEP_control. These effects were weighted by their sample size, N_treatment × N_control/(N_treatment + N_control), where N_treatment and N_control represent the sample size of NEP in warming and control treatment, respectively. We analyzed the relationship between warming-induced changes in NEP and ambient precipitation with a linear mixed-effect model using the lme4 package (39) in R 3.4.1 for Windows. Precipitation was considered as fixed effect, and studies were considered as random effect to interpret possible autocorrelation among observations in each experiment. Moreover, we also used a linear mixed-effect model to analyze the interactions of precipitation, temperature, warming
magnitude, and experimental duration on NEP change (table S2). Similarly, precipitation, temperature, warming magnitude, and experimental duration were considered as fixed effect with a random effect of studies.

We used three ways to justify that MAP is an adequate proxy for soil moisture. First, by substituting SWC with precipitation, we verified that the water-scaling patterns still existed firmly in this study. The warming effects on GEP, ER, and NEP varied from negative to positive when monthly precipitation changed from low to high in our site-level experiment (fig. S5). Second, we used data from a precipitation gradient experiment, which is adjacent to our warming experiment (less than 10 m) (40). We found that precipitation was an adequate proxy for SWC at our study site (fig. S6). Third, at the global scale, previous studies indicate that MAP is a good proxy for soil moisture across large space or various experiments (25).

Statistical analyses
Repeated-measures analysis of variance (ANOVA) was used to explore the effects of warming, year, and their interactions on GEP, ER, and NEP (table S3) and those of measurement time, warming, and their interactions on C fluxes over the growing season for each year (table S4). On the basis of the relationship of GEP, ER, or NEP with ST (fig. S4) (6), we used an exponential function to describe their relationships.

$$F_c(ST) = e^{\gamma_1(ST)}$$  \hspace{1cm} (1)

A quadratic function was used to analyze the relationships between these C fluxes and SWC.

$$F_c(SWC) = \gamma_1 SWC + \gamma_2 SWC^2$$  \hspace{1cm} (2)

For the relationships between C fluxes and the interaction of ST and SWC, both linear and nonlinear models were tried and compared. The linear model assumed that GEP, ER, and NEP were linearly dependent on the SWC.

$$F_c(SWC, ST) = (\gamma_1 SWC + \gamma_2 SWC^2) \times e^{\gamma_1(ST)}$$  \hspace{1cm} (3)

The nonlinear model assumed that ecosystem C fluxes depended on the product of two terms: (i) an increasing exponential function of ST and (ii) a threshold quadratic function of SWC.

$$F_c(SWC, ST) = (\gamma_1 SWC + \gamma_2 SWC^2) \times e^{\gamma_1(ST)}$$  \hspace{1cm} (4)

where $F_c$ represents GEP, ER, or NEP, ST and SWC represent ST and SWC, respectively, and $\gamma$s are the fitted model coefficients.

We performed model selection based on the Akaike information criterion (AIC). Because the nonlinear model (Eq. 4) consistently gave lower AIC (table S5), we selected the nonlinear threshold model, as shown in Eq. 4. The quadratic coefficients ($\gamma$) for SWC were significantly different from zero for GEP, ER, and NEP (table S6), indicating bell-shaped response patterns. We validated the nonlinear model (Eq. 4) by comparing the observed and modeled values. The comparison showed that the model fitted the observations well, with a 49 to 76% agreement for different C fluxes (fig. S7).

We were aware that seasonality might bias the relationships between SWC and GEP, ER, or NEP. To evaluate the potential impacts of this bias, we used peak growing season (July and August) data and performed the same analyses. The results showed the same threshold response of C fluxes to SWC (fig. S8), which further justified the non-linear threshold responses of ecosystem C fluxes to SWC.

Then, we used the model (Eq. 4) to distinguish the impacts of warming-induced changes in ST, SWC, and their interactions on GEP, ER, and NEP (Fig. 4). On average, ST was around 15°C in the control across the 3 years under the current study. Thus, we set 15°C as the baseline ambient ST $[ST(15)]$ and modeled the effects of warming-induced changes in ST and SWC on GEP, ER, and NEP at different SWC levels with 5% intervals (10, 15, 20, 25, 30, 35, 40, 45, and 50%). At each SWC $[SWC(\%)]$ level, the impacts of warming-induced changes in ST were quantified by

$$F_c(SWC(\%), ST(15+\Delta ST)) - F_c(SWC(\%), ST(15))$$  \hspace{1cm} (5)

where $\Delta$SWC represents a warming-induced decrease in SWC.

The combined effects of warming-induced changes in ST and SWC were calculated as

$$F_c(SWC(\%+\Delta SWC), ST(15+\Delta ST)) - F_c(SWC(\%), ST(15))$$  \hspace{1cm} (6)

where $\Delta$ST and $\Delta$SWC were, on average, the differences in ST and SWC of each plot between warming and control treatment across 3 years.

SUPPLEMENTARY MATERIALS
Supplementary material for this article is available at http://advances.sciencemag.org/cgi/content/full/5/8/eaav1131/DC1
Table S1. Site information in global meta-analysis.
Table S2. Effects of mean annual precipitation (MAP), mean annual temperature (MAT), experimental warming magnitude, duration, and their interactions on relative changes in NEP across global herbaceous ecosystems with a linear mixed-effect model.
Table S3. Repeated-measures ANOVA results (F values) on the effects of warming (W), year (Y), and their interactions on GEP, ER, NEP, ST ($T_{50}$), and moisture ($M_{50}$).
Table S4. Repeated-measures ANOVA results (F values) on the effects of warming (W), measured time (T), and their interactions on GEP, ER, NEP, ST ($T_{50}$), and moisture ($M_{50}$) in 2014–2016.
Table S5. Comparison of the nonthreshold and threshold models based on the AIC for GEP, ER, and NEP.
Table S6. Coefficients of threshold model (means and 95% confidence intervals) (Eq. 4).
Fig. S1. Seasonal dynamics and means of ST and SWC at 10-cm depth under three warming treatments in 2014–2016.
Fig. S2. Seasonal means of GEP, ER, and NEP under different warming treatments in 2014 to 2016.
Fig. S3. Warming-induced changes in GEP, ER, and NEP within the year.
Fig. S4. Relationships between ST and GEP, ER, or NEP across seasons and plots.
Fig. S5. Relationships between warming-induced changes in ST and SWC within the year.
Fig. S6. The relationship between response ratio of monthly mean SWC and monthly precipitation in a precipitation gradient experiment at our study site from 2015 to 2016.
Fig. S7. Relationships between warming-induced changes in SWC, GEP, ER, and NEP at different SWC levels with 5% intervals (10, 15, 20, 25, 30, 35, 40, 45, and 50%).
Fig. S8. Relationships between SWC and ecosystem C fluxes within peak growing seasons (July and August).
References (41–65)
REFERENCES AND NOTES


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