Biophysical Controls on Soil Respiration in the Dominant Patch Types of an Old-Growth, Mixed-Conifer Forest

Siyen Ma, Jiquan Chen, John R. Butnor, Malcolm North, Eugénie S. Euskirchen, and Brian Oakley

Abstract: Little is known about biophysical controls on soil respiration in California’s Sierra Nevada old-growth, mixed-conifer forests. Using portable and automated soil respiration sampling units, we measured soil respiration rate (SRR) in three dominant patch types: closed canopy (CC), ceanothus-dominated patches (CECO), and open canopy (OC). SRR varied significantly among the patch types, ranging from 2.0 to 4.5 \( \mu \text{mol m}^{-2} \text{s}^{-1} \) and from 0.9 to 2.9 \( \mu \text{mol m}^{-2} \text{s}^{-1} \) during the 1999 and 2000 measuring periods, respectively, with the maximum in CECO and the minimum in OC. Multiple peaks of seasonal SRR were functions of soil temperature and moisture dynamics. The relationship between SRR and soil temperature switched from a positive to a negative correlation when soil moisture dropped from saturation to drought. Time lag, as a function of soil moisture, was included in an exponential model to assess the effects of soil moisture on SRR in this seasonal water-stressed ecosystem. The total soil C flux summed by an area-weighted average across all three patch types was 660 \( \pm 163 \) g C m\(^{-2}\) from May to Oct. 2000. These results may be applicable to other water-stressed forests in the Mediterranean climate zone, and have implications for the conservation of soil carbon. For. Sci. 51(3):221–232.

Key Words: Soil CO\(_2\) efflux, soil temperature, soil moisture, Sierra Nevada, California.

Soil CO\(_2\) efflux is a critical component of the terrestrial CO\(_2\) cycle under changing climatic conditions. The terrestrial ecosystem releases 50–70 Pg C yr\(^{-1}\) from the soil to the atmosphere, over 10 times the quantity of CO\(_2\) released through combustion of fossil fuels (Raich and Schlesinger 1992). Although forests play an important role in C uptake through photosynthesis, they also release C via autotrophic and heterotrophic respiration (Singh and Gupta 1977, Raich and Schlesinger 1992). Soil respiration is related to many ecological processes, such as photosynthesis, root respiration, organic matter decomposition, and microbial activity (Sørensen 1974, Bunnell et al. 1977, Orchard and Cook 1983, O’Connell 1990, Ryan et al. 1996, Mallik and Hu 1997, Lomander et al. 1998, Hart and Sollins 1998, Högberg et al. 2001). These processes are interrelated and affected by multiple biophysical factors in an ecosystem. Soil respiration is also associated with physical CO\(_2\) diffusion processes affected by physical properties of soil and litter. Interpretation and prediction of soil respiration depends strongly upon ecosystem type, and it is difficult to apply one model consistently across a variety of ecosystems. For example, the \( Q_{10} \) model has been used in several ecosystem models, but the \( Q_{10} \) value is not necessarily a constant (Rustad et al. 2001, Xu and Qi 2001, Janssens and Pilegaard 2003) and can bias soil respiration estimation.

Numerous efforts have been made to understand the mechanisms behind the fluctuation of soil respiration and develop empirical models to predict soil respiration from biophysical factors, such as soil temperature and moisture (Singh and Gupta 1977, Schlentner and Van Cleve 1985, Gordon et al. 1987, Zimov et al. 1993, Paivari 1995, Simmons et al. 1996, Billings et al. 1998, Davidson et al. 1998, Lomander et al. 1998, Russell and Voroney 1998, Law et al. 2000, Maier and Kress 2000, Xu and Qi 2001, Liu et al. 2002, Euskirchen et al. 2003). Although a positive relationship has been reported between soil respiration and temperature at many sites (Lloyd and Taylor 1994), the relationship can be very weak or even negative when soil moisture is limited (Bunnell et al. 1977, Xu and Qi 2001). Such negative correlations between soil respiration and soil temperature have been observed in Mediterranean climate zones (Wildung et al. 1975, Carlyle and Ba Than 1988, Luo et al. 1996, Wang et al. 1999, Casals et al. 2000, Ma et al. 2004). A key question is what are the mechanisms that...
determine this inverse relationship? Examining specific biophysical controls on soil respiration in these ecosystems is needed to address this question.

Unlike many temperate forests, California’s Sierra Nevada mixed-conifer forest has readily delineated patches with distinct microclimate, soil, and vegetation characteristics. The forest experiences a typical Mediterranean climate of hot, dry summers and cold, wet winters. The summer drought conditions constrict aboveground and belowground processes (Schimel et al. 1999, Law et al. 2000, Royce et al. 2001a and b, Chen et al. 2002). Recent research has emphasized the significant influence of carbon assimilation on soil respiration and the importance of vegetation conditions in understanding soil respiration patterns (Högberg et al. 2001, Giardina and Ryan 2002, Giardina et al. 2003), but the combination of temperature and soil moisture in seasons may profoundly influence not only the aboveground but also belowground processes. We conducted this research in the heterogeneous mixed-conifer forest to gain more complete understanding of the vegetation distribution and biophysical controls on soil respiration, and to develop empirical models for estimating soil respiration that might be applied to other water-stressed ecosystems. Specifically, the objectives of this study were to (1) examine the differences of soil respiration between dominant patch types; (2) investigate temporal variations in soil respiration and its relationship to soil temperature and moisture; and (3) estimate annual soil carbon efflux using empirical models based on field measurements.

Methods

Study Site

Our study was conducted in the Teakettle Experimental Forest (TEF, 36°58’ N, 119°2’ W) located at 1,880–2,485 m elevation in the Sierra National Forest, California. The average annual precipitation at the TEF is 1,250 mm, but most of it falls during intense winter snowstorms between Nov. and Apr. The depth of the winter snowpack is an important factor influencing the soil water conditions of the forest during the following summer. The mean air temperature in July is 14.5° C, whereas the mean air temperature in Jan. is 10.0° C. Soils are generally Xerumbrepts and Xeropsamments typical of southwestern slopes of the Sierra Nevada. The granite-based soils have a coarse sandy loam texture throughout the relatively shallow profile (75–100 cm), with ~18–20% volume soil water-holding capacity. Soil bulk density at 20–25 cm depth is 1.09 g cm⁻³. Duff and litter comprise 90.5% of soil cover, bare soil 5.8%, and large woody debris 3.7%. Organic matter content within the 0–10-cm depth of the mineral soil is 6.4% (North et al. 2002 and references therein).

In the TEF old-growth mixed-conifer forest, the major conifer species include white fir (Abies concolor), sugar pine (Pinus lambertiana), incense cedar (Calocedrus decurrens), Jeffrey pine (Pinus jeffreyi), and red fir (Abies magnifica). The mixed-conifer forest is a matrix of tree clusters punctuated with gaps averaging 5–20 m in diameter. Gaps are occupied by bare ground and several shrubs, particularly Ceanothus cordulatus. Previous research at TEF investigated the percent cover of vegetation, canopy cover, coarse woody debris (CWD, >10-cm diameter), bare ground, and rock at 402 grid-sampling points. At each point, vegetation cover (including herbs, shrubs, and seedlings) was visually estimated after calibration with a standard cover size, i.e., one sheet of regular paper = 1.5% cover of 10-m² plots. Canopy cover was measured using a moosehorn (a funnel with a bubble level, a 30° angle of view off vertical, and a transparent grid with 100 intersections affixed over the top of the funnel). Litter depth (cm) was also measured by averaging three separate measurements from the top of the organic layer to the mineral soil. All the information was grouped with hierarchical clustering using Ward’s method and a relative Euclidean distance measurement (McCune and Grace 2002). Pruning of the dendrogram to establish the optimal number of clusters was done using Indicator Species Analysis (Dufrene and Legendre 1997). Three patch types were classified, including closed canopy tree groups (CC), patches of whitehorn ceanothus (Ceanothus cordulatus) (CECO), and open canopy (OC). The CC, CECO, and OC occupy 67.7, 13.4, and 4.7% of the entire forest, respectively, with most of the remainder composed of exposed granite (North et al. 2002).

Soil Respiration Measurements

Soil respiration rates (SRR, µmol m⁻² s⁻¹) were measured using an EGM-2 Environmental Gas Monitor equipped with a SRC-1 Soil Respiration Chamber (PP Systems, Hitchin, Herts, UK) and an automatic soil-respiration measuring system, the Automatic Carbon Efflux System (ACES, US patent no. 6,692,970, Butnor 1999, Butnor et al. 2003). Both systems were used to measure SRR from dynamic chambers and analyzed CO₂ concentration with the same model of infrared gas analyzer (EGM-2 Environmental Gas Monitor, PP Systems). However, the two systems used different chamber designs and measurement techniques. The PP Systems SRC-1 has a tendency to overestimate SRR on sites that have soil (or litter) with high air content or porosity, whereas the ACES has an integrated calibration that is accurate regardless of soil or litter condition (Le Dantec et al. 1999). We took parallel measurements using these two units and established a linear equation (SRR_correct = 0.467 SRR_SRC −1.136 + 0.123, R² = 0.91) to correct the PP systems SRC-1 biases. The two sets of EGM-2 infrared gas analyzers were calibrated using a standard CO₂ gas (700.1 ppm) under actual air pressure once a week.

The SRC-1 uses a smaller chamber (78.5 cm²) with a “closed” gas-sampling design. It is portable and therefore was used to measure SRR rapidly over an extensive area. We used an SRC-1 unit and measured daytime (i.e., between 0900 and 1600 local hours) soil respiration once a week during June-Aug. 1999 and once another week or once a month during May–Oct. 2000. In 1999, we randomly selected six replications from each of the three patch types.
In 2000, we increased the total sample size from 18 to 64, and expanded the sample area. The replications for CC, CECO, and OC were 27, 20, and 17, respectively. At each sampling point, four 10-cm diameter PVC collars were installed several days before taking SRR measurements to allow for settling. Air pressure was measured with a pocket barometer to correct the original field data using company-recommended equations (PP Systems EGM-2 manual 1997).

The ACES uses larger steady-state soil chambers (490.8 cm²) with an “open” gas-sampling design, i.e., the chambers have pressure-equilibration ports and are always in direct communication with atmospheric pressure. The unit can be left in place for extended periods to detect subtle changes in SRR over time. In this study, the ACES chambers covered the sampling area during one sampling period, typically 2–3 days, and then the chambers were opened for maintenance purposes. If it rained, the chambers were moved to next-wet spots right after the rain. Fifteen chambers were connected to a central manifold and gas analyzer via flexible tubing (15.5 m) and were measured sequentially. Because it took 10 min for each chamber to come into equilibrium and complete a single measurement, nine measurements were taken at each of the fifteen chambers during a 24-hour cycle (i.e., with intervals of 2.667 h). When the chambers were not actively being sampled, they were continuously refreshed by an exhaust pump system. Multiple 12-V marine batteries were connected in parallel to power the ACES. The batteries were replaced and recharged every 3–4 days to assure steady power supplies. An ACES unit was located randomly at three areas from May 14 to June 12, from June 13 to June 30, and from July 1 to July 9.

To measure the CO₂ concentration in the soil profile, we connected an aquarium air stone to a plastic tube and inserted it into the soil. The middle of the air stone was inserted to depths of 5, 10, and 20 cm from the surface, and the other end of the tube was connected to the EGM-2 “Gas In” connector. CO₂ concentration was recorded when the reading became stable. Because the flow rate of EGM-2 is low (~300 ml min⁻¹), we assumed that there was no significant influence on the CO₂ profile during a measuring period less than one minute. We placed a plastic cap on the top of the tube in the ground to prevent gas leaking through the tube. CO₂ concentrations were measured weekly at 18 sampling points.

**Environmental Variable Measurements**

While measuring SRR, we simultaneously recorded soil temperature at 5-cm depth (Tₛ, °C) and volumetric soil moisture within the 0–15-cm depth (Mₛ, %). For the SRC-1, Tₛ was measured using a digital thermometer (Taylor Digital Max/Min, Forestry Suppliers, Inc., Jackson, MS). Tₛ associated with ACES was measured using thermocouple probes installed inside each chamber (Butnor 1999). A Time Domain Reflectometry unit (TDR, Model 6050XI Soil moisture Equipment Corporation, Santa Barbara, CA) was used to measure Mₛ within 0–15-cm depth at each of the SRR measuring locations for both systems (Gray and Spies 1995). Two 30-cm-long stainless steel rods were inserted into the mineral soil at a 30° angle to the surface to increase the sensitivity of TDR measurements in parallel. We performed a calibration function and calculated soil water volumetric content (θₛ, %) based on the TDR reading (Δt, ns): θₛ = −10.16 + 10.34 × Δt (Gray and Spies 1995). TDR measurements were taken from probes located within 10 cm of each ACES chamber once a week throughout the measuring period. The Mₛ readings associated with ACES were converted to soil water volumetric content using a predefined function. Soil surface temperature (Tₛ, °C) and soil water potential at 15-cm depth (SWP, kPa) were continuously measured using Model 257 Soil Moisture Sensors connected to a CR10 datalogger (Campbell Scientific, Inc. Logan, UT).

**Data Analysis**

Analysis of variance (ANOVA) was used to test significant SRR differences among three dominant patch types within each of the sampling years (i.e., 1999 and 2000). We also used ANOVA to test the significant effects of canopy cover, CWD cover, ceanothus cover, and litter depth on soil respiration variation. These analyses were performed using the General Linear Model for unbalanced data sets (Cody and Smith 1997). We calculated Pearson correlation coefficients (r) between SRR and time-lagged Tₛ in a 24-h period. Linear regression analysis was applied to establish the exponential model between SRR and Tₛ coupling with Mₛ and time lag in different patch types.

\[
SRR = a \cdot e^{\beta T_{s5}(\text{Lag})},
\]

where α and β are coefficients estimated by the general linear regression model. Lag, is the time lag between SRR and Tₛ, which is determined by Mₛ. One unit of lag time is equivalent to 2.667 h associated with the ACES system design. Mean Tₛ for 2.667-h intervals was calculated from soil surface temperature (Tₛ, °C) based on the following linear regression models developed for our study site (Tₛ = 0.7741Tₛ + 3.15, R² = 0.95). Soil water potential (SWP) was converted to Mₛ based on an empirical soil–water characteristic curve established by regressing the paired Mₛ and SWP data at the same sampling point in the field (Mₛ = 7.9244 SWP⁻⁰.₃₄¹⁷, R² = 0.94). Based on the model, SRR was estimated for each patch type, and then soil carbon flux for the entire forest was summarized from the weighted contribution of each patch type to the total area. All analyses were performed with SAS (V8.0, SAS Institute Inc., Cary, NC) with a significance level of 0.05.

**Results**

**SRR in the Three Patch Types**

The mean SRR using the SRC-1 unit was significantly different among the three patch types (F = 35.17, P < 0.0001), ranging from 2.0 to 4.5 μmol m⁻² s⁻¹ and from
Changes over time

SRR fluctuated over time and responded to changes of soil temperature and moisture. Temporal SRR patterns, measured with the ACES unit, appeared to be more dynamic in the CC and CECO patch types than in the OC patch type during the 29-day period from May 14 to June 12 in 2000 (Figure 3, a–c). Temporal SRR patterns in the CC and CECO patches followed changes in $T_{s5}$ measured at the same time but were confounded by changes in SWP (Figure 3d). Low SRR between days 135 and 142 was coupled with low $T_{s5}$ (slightly above 0°C), which was likely related to a snowstorm (12–31 cm deep) on May 15, 2000 (day 135). However, SRR was low between days 156 and 160 while $T_{s5}$ was high, and high SRR was measured while $T_{s5}$ was low after day 160. A significantly elevated SWP resulted from the precipitation on day 156. The responses of SRR to the change of $T_{s5}$ varied before and after the rain event.

SRR had clear diurnal patterns in the three patch types, and these patterns varied greatly throughout the season (Figure 3, a–c). During the measurement period, the greatest daily range of SRR generally occurred in CECO. Compared to CC and CECO, OC had the smallest range in daily pattern, but this was coupled with the largest daily temperature range. The diurnal pattern did not appear to be influenced by $M_s$, because $M_s$ was influenced by precipitation events and varied at a relatively larger temporal scale.

Empirical Models

Seasonally, the relationship between SRR and temperature switched from positive to negative, and changes in $M_s$ were associated with changes in the degree and direction of the correlation (Figure 5). SRR was positively correlated to $T_{s5}$ when $M_s$ was higher than 10% (Figure 5a) in all three patch types, with an $R^2$ value of 0.79, 0.43, 0.48 in the OC, CC, and CECO patches, respectively. These positive relationships were weakened when $M_s$ decreased to 5–10% (Figure 5b) and were negative for the CC and CECO patch types when $M_s$ was ≤5% (Figure 5c). The peaks in r values between SRR and time-lagged $T_{s5}$ indicate that SRR was positively correlated to the current $T_{s5}$ at any time of the day regardless of the patch type when soil moisture was high. When soil became dry, SRR was negatively correlated to current $T_{s5}$ but positively correlated to the temperature that occurred 5–10 h before SRR measurements (Figure 6).

On the basis of this finding, we modeled SRR changes to $T_{s5}$ time-lagged by three $M_s$ levels. The confounding variance of patch type, $T_{s5}$, and $M_s$ accounted for 52–91% of the SRR variation (Table 2). The empirical model estimated that cumulative SRR was 372, 819, and 716 g C m$^{-2}$ in OC, CC, and CECO, respectively, during the measuring period from May to Oct. 2000. Using these cumulative SRRs and the percent cover of each patch type, we calculated the weighted average soil C flux as 660 ± 163 g C m$^{-2}$ between May and Oct. 2000 in this mixed-conifer forest.

Discussion

The finding that soil respiration is significantly different among patch types supports other research suggesting the importance of vegetation, roots, and microbial communities in understanding variable soil respiration patterns within heterogeneous forests (Buchmann 2000, Raich and Tufekcioglu 2000). Patch type may explain the differences in soil respiration from three perspectives: root respiration, rhizospheric microbial respiration, and free-living microbial respiration (Kelting et al. 1998). Hanson et al. (2000) concluded that the amount that root respiration contributes to total soil respiration depends on ecosystem type and seasonal pattern. In the TEF, fine-root biomass is greater in CC than in CECO and OC within the top 10 cm of soil. Higher
soil respiration in CC may result from greater fine-root biomass than in OC. OC patches probably have few fine roots, and heterotrophic respiration from the mineral soil is low (Kelting et al. 1998). Nevertheless, root biomass does not account for all the patterns we measured. The highest SRR in CECO may be the result of the presence of the symbiont *Frankia* spp., a type of N-fixing actinomycete that provides high nitrogen availability in the ceanothus rhizosphere (Oakley et al. 2003). Symbiotic and free-living microbes in the rhizosphere likely play an important role in SRR in CECO. Unfortunately, there is no data for directly comparing the physiological processes and seasonal dynamics between trees and shrubs, but tree growth has been shown to significantly slow during the summer drought in the Sierra Nevada (Royce and Barbour 2001b). Consequently, root respiration may contribute relatively less to soil respiration than heterotrophic respiration during the water-stress season (Hanson et al. 2000). Water stress would limit photosynthesis and reduce current carbon assimilation, resulting in a decrease in living root respiration (Högberg et al. 2001, Shah and Paulsen 2003). The significant association between SRR and litter depth and CECO cover implies that heterotrophic respiration may become relatively more important than root respiration during the dry season. Soil respiration generally increased with increases in litter depth (Figure 2d), suggesting that aboveground productivity influenced soil respiration (Euskirchen et al. 2003). Further studies are needed to understand the decrease in SRR associated with litter deeper than 10 cm. We also noticed that the relationship between CWD and SRR was nearly parabolic,
perhaps indicating that microbial community structure, the biomass, or activities are different in different substrate natures of litter and CWD (Bunnell et al. 1977). Belowground processes, such as decomposition of the litter and organic matter, are critical to future investigations of SRR in water-stressed ecosystems.

We organized our data from the biophysical perspective, but the diurnal and seasonal patterns of soil respiration reflect changes in the related ecological processes and their interactions with the biophysical environment. In mesic ecosystems, soil respiration usually reaches its maximum between noon and late afternoon, and the minimum occurs
between midnight and early morning (Singh and Gupta 1977 and references therein). In these ecosystems, soil respiration is clearly regulated by the diurnal patterns of photosynthesis. However, in water-stressed ecosystems, the diurnal patterns of soil respiration appear to be different. We graphed Figure 5 using time of day on the horizontal axis and found that the maximum rate of soil respiration occurred at night, whereas the minimum occurred between noon and late afternoon. Microbial respiration involves physiological and biochemical processes, which differ from those associated with root respiration. Because microbes have different sensitivity to changes in the biophysical environment (Boone et al. 1998), these unusual diurnal patterns reinforce the notion that heterotrophic respiration may be more important than root respiration when soil moisture limits photosynthesis (Singh and Gupta 1977 and references therein). Thus, soil moisture is a critical biophysical factor controlling the magnitude and pattern of soil respiration in water-stressed ecosystems.

In this study, the seasonal patterns of soil respiration differ from the bell-shaped seasonal pattern reported in temperate and boreal forests, where soil water does not limit plant physiological processes during the growing season (Schlentner and Van Cleve 1985, Gordon et al. 1987, Zimov et al. 1993, Pajari 1995, Simmons et al. 1996, Billings et al. 1998, Davidson et al. 1998, Russell and Voroney 1998, Maier and Kress 2000). In our study site, high soil moisture usually occurs during the period immediately after snowmelt (early May) and ends at the beginning of the summer drought (the end of June). During this “wet” season, air temperature increases to over 10° C, and rapid vegetation growth is associated with relatively high temperature and moisture availability (Royce and Barbour 2001b). During this period, soil respiration reaches its seasonal peak and may be largely governed by the vegetation community. After the snowmelt, soil moisture dramatically declines and quickly reaches a point where plant physiological processes become stressed, resulting in decreased root respiration (Högberg et al. 2001, Shah and Paulsen 2003). At times when soil moisture may be too low to be used by plants, there may still be sufficient moisture for microbial communities, because microbes respond differently to changing biophysical environments (Boone et al. 1998). Therefore, soil respiration is more driven by microbial respiration and remains relatively stable (Giardina and Ryan 2000).

Soil respiration is also related to a physical process, CO2 diffusion, especially when rain events occur (Cook et al. 1998, Bouma and Bryla 2000). High CO2 concentration in the soil profile generally indicates high soil respiration rates. However, the current chamber technique may not be able to detect the actual increase in soil profile. For example, we notice that SRR immediately decreases after the rain regardless of soil temperature, gradually returning to its prerain level, and occasionally remaining elevated, perhaps because of the reduction of CO2 diffusion—the blocking effect of rain event (Figure 3 and 4). Bouma and Bryla (2000) reported that soil CO2 concentration increased after irrigation, but CO2 efflux on the soil surface decreased. This finding suggests that CO2 may not be released to the air from the soil surface immediately despite high belowground CO2 concentrations from moisture-driven root and microbial respiration. Thus, soil CO2 profile measurement is more proper to examine the biological processes under the ground. In contrast, O2 is necessary for the respiration process of roots, many microbes, and soil fauna. As soil water content increases, soil aeration and O2 availability decreases, and soil respiration decreases (Vor et al. 2003). With high soil moisture, elevated CO2 production from root and microorganisms can produce a high concentration of CO2 in the soil, which can reduce root and microbial activity accordingly. In turn, SRR after a lasting rain event may be greatly reduced. The blocking effect related to rainfall in arid ecosystems may involve several key soil processes that...
have received little attention in soil C flux studies. Con-
trolled experiments that measure CO₂ concentration in the
soil profile, soil respiration, microclimate, and biological
activities simultaneously are needed to understand and
model such complex interactions (Billings et al. 1998).

Although most temperate forests have a positive relation-
ship between SRR and temperature, at TEF the relation-
ship changed from positive to negative with the onset of the
summer drought (Figure 5). Our study found that the posi-
tive relationship did exist during the wet season. Under this
condition, autotrophic respiration may regulate soil respira-
tion more strongly than heterotrophic respiration, and soil

* Table 2. Discrete models relating soil temperature at 5-cm depth ($T_{s5}$, °C) and at three soil moisture ($M_s$, %) levels to predict soil respiration rate (SRR, $\mu$mol m$^{-2}$ s$^{-1}$)

<table>
<thead>
<tr>
<th>Model</th>
<th>$M_s$</th>
<th>Patch Type</th>
<th>$\alpha$</th>
<th>$\beta$</th>
<th>Lag</th>
<th>$R^2$</th>
<th>$P &gt; F$</th>
<th>$n$</th>
</tr>
</thead>
<tbody>
<tr>
<td>SRR = $ae^{\beta(T_{s5})}$</td>
<td>&gt;10.0</td>
<td>CC</td>
<td>3.109</td>
<td>0.059</td>
<td>0</td>
<td>0.91</td>
<td>&lt;0.0001</td>
<td>780</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CECO</td>
<td>3.723</td>
<td>0.531</td>
<td></td>
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<td></td>
<td></td>
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<tr>
<td></td>
<td></td>
<td>OC</td>
<td>2.271</td>
<td>0.531</td>
<td></td>
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<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>5.0–10.0</td>
<td>CC</td>
<td>3.774</td>
<td>0.047</td>
<td>5</td>
<td>0.89</td>
<td>&lt;0.0001</td>
<td>267</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CECO</td>
<td>2.721</td>
<td>0.531</td>
<td></td>
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<td></td>
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</tr>
<tr>
<td></td>
<td></td>
<td>OC</td>
<td>1.882</td>
<td>0.531</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>&lt;5.0</td>
<td>CC</td>
<td>2.076</td>
<td>0.059</td>
<td>10</td>
<td>0.52</td>
<td>&lt;0.0001</td>
<td>168</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CECO</td>
<td>2.039</td>
<td>0.531</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td>OC</td>
<td>1.050</td>
<td>0.531</td>
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</tr>
</tbody>
</table>

The fit parameters ($\alpha$ and $\beta$) are for the performance of the model across all three patch types (i.e., CC, canopy cover; CECO, ceanothus shrub; and OC, open canopy) at a given $M_s$ range and time lags (Lag, in hours).

Temperature is a good predictor of soil respiration (Qi and Xu 2001). However, the seasonal switch to a negative relationship may indicate that the dominant ecological pro-
cesses are switched from autotrophic root to heterotrophic microbial processes. This switch should be considered when modeling soil respiration across a variety of ecosystems that may include water-stressed communities at broad scales.

Our empirical model combined soil moisture influences with a time lag of soil temperature, and its estimation of the total soil C flux for the old-growth mixed conifer forest is well within the range for old-growth ponderosa pine forests (Irvine and Law 2002) and other Mediterranean forests
(Raich and Schlesinger 1992). Although the temperature time lag is included in our model, more field data and controlled experiments are needed to understand the underlying processes that influence the time lag. For example, assimilated C needs several days to transfer from canopy to root and rhizosphere (Ekblad and Höglberg 2001). Time lag may reflect the responses of C assimilation and allocation to the combination of soil temperature and moisture. From a modeling perspective, soil temperature at greater depth (>5 cm) may take the place of the time lag because the maximum temperature at greater depth is later in the day. The time lag between the maximum surface and soil temperature at 30–40-cm depth can be as much as 10 hrs (Campbell and Norman 1998). At TEF, however, we found that <20% of the coarse roots are distributed in the top 20 cm of soil, whereas nearly 60% are found in the 20–60-cm layer. Although, near-surface temperatures might not be as well correlated with soil respiration as deeper soil temperatures (Irvine and Law 2002). More studies are needed to examine the spatial distribution of roots and microbial biomass and the correlation of their distribution with soil temperature in the different patch types.

In water-stressed forests, it is difficult to build an empirical model to predict soil respiration solely based on soil temperature and moisture. We introduced the “patch type” variable into the empirical model because patch type combines both the vegetation and the microbial communities. Predicting soil respiration with biophysical factors is a convenient approach to describe complex ecosystem processes using a few simple and easily obtained environmental variables. \( Q_{10} \), the rate increase in a process with a 10°C increase in temperature, indicates the sensitivity of SRR to soil temperature (Lloyd and Taylor 1994). SRR can be modeled by the \( Q_{10} \)-based model as \( SRR = SRR_0 Q_{10}^T \), where \( SRR_0 \) represents a soil respiration rate at a standard temperature and where \( T \) is soil temperature. The exponential model (\( SRR = Ae^{BT} \)) has also been widely used to predict SRR from temperature (Lloyd and Taylor 1994, Luo et al. 2001, Qi et al. 2002). The \( Q_{10} \)-based model can be transformed to the exponential model using \( Q_{10} = e^{10B} \) (Davidson et al. 1998). Considering the relation between soil moisture and time lag, we computed the range of \( Q_{10} \) from 1.6 to 1.9 among the three patch types in our study site, indicating the temperature sensitivity of soil respiration. However, when time lag is not considered, the \( Q_{10} \) values have higher fluctuation associated with variation in litter depth (Ma et al. 2004). This variability may indicate that surface communities are less sensitive to soil temperature but more sensitive to moisture than the deeper communities (Bunnell et al. 1977, Casals et al. 2000). Climatic dynamics, vegetation, soil types, and nutrient conditions may all cause large ranges in \( Q_{10} \) values (Raich and Schlesinger 1992, Luo et al. 1996, Davidson et al. 1998, Lomander et al. 1998, Gulledge and Schimel 2000, Qi and Xu 2001). Using the exponential model to estimate soil respiration from ecosystem to global scales may be better than using the \( Q_{10} \)-based model because the parameter (B) in the exponential model can reflect \( Q_{10} \) variations (Qi et al. 2002). The relation between the \( Q_{10} \)-based model and the exponential model can improve our understanding of the relationship between soil respiration and biophysical controls.

![Figure 6. Correlation coefficients (r) between soil respiration rate and soil temperature at 5-cm depth (\( T_m \)) with each 2.5-hour lag time in three patch types: (a) open canopy (OC), (b) closed canopy (CC), and (c) ceanothus shrub (CECO) at three soil moisture (\( M_s \)) conditions: I: \( M_s > 10\% \); II: 5% \(< \ M_s \leq 10\% \); and III: \( M_s \leq 5\% \).](image-url)
Many questions remain to be answered regarding the effects on soil C fluxes of natural and anthropogenic disturbance in a changing climate. We hope this research emphasizing the importance of vegetation patch and soil moisture conditions on SRR can help develop alternative management plans and monitoring yardsticks for conservation of soil carbon.

**Literature Cited**


BUTNOR, J.R., K.H. JOHNSEN, R. OREN, AND G.G. KATUL. 2003. Many questions remain to be answered regarding the effects on soil C fluxes of natural and anthropogenic disturbance in a changing climate. We hope this research emphasizing the importance of vegetation patch and soil moisture conditions on SRR can help develop alternative management plans and monitoring yardsticks for conservation of soil carbon.

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