Modeling the effects of snowpack on heterotrophic respiration across northern temperate and high latitude regions: Comparison with measurements of atmospheric carbon dioxide in high latitudes


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Abstract. Simulations by global terrestrial biogeochemical models (TBMs) consistently underestimate the concentration of atmospheric carbon dioxide (CO2) at high latitude monitoring stations during the nongrowing season. We hypothesized that heterotrophic respiration is underestimated during the nongrowing season primarily because TBMs do not generally consider the insulative effects of snowpack on soil temperature. To evaluate this hypothesis, we compared the performance of baseline and modified versions of three TBMs in simulating the seasonal cycle of atmospheric CO2 at high latitude CO2 monitoring stations; the modified version maintained soil temperature at 0 °C when modeled snowpack was present. The three TBMs include the Carnegie-Ames-Stanford Approach (CASA), Century, and the Terrestrial Ecosystem Model (TEM). In comparison with the baseline simulation of each model, the snowpack simulations caused higher releases of CO2 between November and March and greater uptake of CO2 between June and August for latitudes north of 30° N. We coupled the monthly estimates of CO2 exchange, the seasonal carbon dioxide flux fields generated by the HAMOCC3 seasonal ocean carbon cycle model, and fossil fuel source fields derived from standard sources to the three-dimensional atmospheric transport model TM2 forced by observed winds to simulate the seasonal cycle of atmospheric CO2 at each of seven high latitude monitoring stations. In comparison to the CO2 concentrations simulated with the baseline fluxes of each TBM, concentrations simulated using the snowpack fluxes are generally in better agreement with observed concentrations between August and March at each of the monitoring stations. Thus, representation of the insulative effects of snowpack in TBMs generally improves simulation of atmospheric CO2 concentrations in high latitudes during both the late growing season and nongrowing season. These simulations highlight the
global importance of biogeochemical processes during the nongrowing season in estimating carbon balance of ecosystems in northern high and temperate latitudes.

Introduction

There is evidence that warming is occurring in some high-latitude areas (Lachenbruch & Marshall 1986; Beltrami & Mareschal 1991; Chapman & Walsh 1993), and that the warming may be impacting both ecosystem function and structure (Oechel et al. 1993, 1995; Chapin et al. 1995). These ecosystems contain approximately 40% of the world’s soil carbon inventory that is potentially reactive in the context of near-term climate change (McGuire et al. 1995; Melillo et al. 1995; McGuire & Hobbie 1997). A substantial amount of carbon could be released in inorganic forms from these soils in response to elevated temperature (Nadelhoffer et al. 1992; Oechel et al. 1993, 1995). A large release of CO₂ from these soils has the potential to influence the growth of atmospheric CO₂, which may have consequences for the rate and magnitude of climate change.

During the late 1980’s, substantial releases of CO₂ were observed from Alaskan tundra ecosystems in response to declining water tables associated with elevated temperature (Oechel et al. 1993). Recent measurements of CO₂ exchange between tundra ecosystems and the atmosphere indicate that substantial losses of CO₂ from tundra soils may occur during the fall, winter, and spring months (Zimov et al. 1993, 1996; Oechel et al. 1997). Recent measurements at some locations indicate that these losses are currently greater than sink activity in summer months (Oechel and Vourlitis, unpublished). Thus, estimates of carbon balance in high latitude ecosystems are potentially biased if they are based on summer measurements alone (Oechel et al. 1997).

A number of global terrestrial biogeochemical models (TBMs) have been developed to assess the effects of changes in climate and atmospheric carbon dioxide on terrestrial ecosystem processes at large spatial scales (Heimann et al. 1998; Cramer et al. 1999; Kicklighter et al. 1999). For typical global application, TBMs make estimates of monthly CO₂ exchanges between terrestrial ecosystems and the atmosphere for approximately 60,000 grid cells at a spatial resolution of 0.5° by 0.5° (latitude by longitude), with about one-third of these estimates in high latitudes. Heimann et al. (1998) compared the performance of several TBMs by using the monthly estimates of CO₂ exchange by the models to simulate the seasonal cycle of atmospheric CO₂ at a number of CO₂ monitoring stations located throughout the globe. At high latitude monitoring stations, the simulations consistently underestimated
the concentration of atmospheric CO₂ during the nongrowing season, which suggests that global TBMs tend to underestimate the release of CO₂ from high latitude soils during the nongrowing season. We hypothesize that decomposition is underestimated during the nongrowing season primarily because the TBMs do not consider the insulative effects of snow on soil temperature. In this study we modify the decomposition formulation of three TBMs to evaluate whether consideration of the insulative effects of snowpack improves the simulation of monthly concentrations of atmospheric CO₂ at CO₂ monitoring stations located in high latitudes.

**Modeling the seasonal cycle of atmospheric CO₂**

**Overview**

To model the seasonal cycle of atmospheric CO₂ at a monitoring station requires spatially explicit seasonal estimates of carbon exchange between the atmosphere, the oceans, and the terrestrial surface (Figure 1). In the absence of disturbance and major climatic fluctuations, annual net primary production (NPP) and annual heterotrophic respiration (R₇; i.e., decomposition) of terrestrial ecosystems are approximately in balance. Because photosynthesis and microbial activity are controlled in different ways by environmental and biotic factors, the magnitudes of NPP and R₇ are not synchronized throughout the year. Here we adopt the ecological convention that fluxes into the land surface are positive. On a monthly basis, net ecosystem production (NEP), which is the difference between monthly NPP and R₇, can either be positive or negative; positive values indicate sink activity and negative values represent source activity. The seasonal fluctuations of NEP are primarily responsible for the observed seasonal variation in the concentration of atmospheric CO₂, especially at high northern latitudes where these variations have the greatest amplitude (Fung et al. 1983, 1987; Heimann et al. 1989, 1998).

In this study we conducted two simulations with each of three TBMs to evaluate how a representation of the insulative effects of snowpack influences the seasonal cycle of atmospheric CO₂ at high latitude monitoring stations in comparison to the baseline version of each model. For the snowpack version of each model, we modified the R₇ formulation in the baseline version so that temperature is maintained at 0 °C when modeled snowpack is present. For each simulation in this study, monthly NPP, R₇, and NEP are estimated for each 0.5° grid cell of the terrestrial biosphere. The spatially explicit monthly NEP estimates are combined with equivalent CO₂ flux fields generated by an ocean biogeochemical model, and fossil fuel source fields; monthly CO₂ emission fields from fossil fuel burning and cement manufac-
ture are computed based on a global 1° × 1° map compiled by Marland et al. (1989) assuming constant emissions throughout the year. The terrestrial, oceanic, and fossil fuel fields of CO₂ fluxes provide the lower boundary condition for a three-dimensional atmospheric transport model forced by observed winds. To evaluate the simulation of atmospheric CO₂ concentrations, we extracted the monthly estimates of atmospheric CO₂ from locations corresponding to monitoring stations and compared the estimates with the detrended observations at each station.

In a previous study (Heimann et al. 1998), we obtained observed atmospheric CO₂ concentrations for the years 1983–1992 from the monitoring program of the National Oceanographic and Atmospheric Administration (NOAA; Conway et al. 1994a, 1994b). The seasonal cycle of atmospheric CO₂ was extracted from the raw station data at 27 monitoring stations, as described by Heimann et al. (1998). Among the 27 stations, we focus our comparison between simulated and observed concentrations of atmospheric CO₂ on the seven most northern stations (Table 1). Because these stations effectively integrate seasonal CO₂ exchanges across much of the northern hemisphere (Kaminski et al. 1996), we focus our evaluation of the seasonal patterns of CO₂ exchange on northern high latitudes between 60° N and 90° N and on northern temperate latitudes between 30° N and 60° N. To evaluate how well a simulation reproduces the observed seasonal signal of atmospheric CO₂ at a monitoring station, we calculated a normalized mean-squared deviation (NMSD) as

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**Figure 1.** Flowchart of the data flow and model linkages for simulating the seasonal cycle of atmospheric CO₂.
Table 1. Stations from the NOAA station network (Conway et al. 1994b) used in this study.

<table>
<thead>
<tr>
<th>#</th>
<th>Abbreviation</th>
<th>Station</th>
<th>Country</th>
<th>Latitude</th>
<th>Longitude</th>
<th>Elevation [m]</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>ALT</td>
<td>Alert, N.W.T.</td>
<td>Canada</td>
<td>82°27' N</td>
<td>62°31' W</td>
<td>210</td>
</tr>
<tr>
<td>2</td>
<td>MBC</td>
<td>Mould Bay, N.W.T.</td>
<td>Canada</td>
<td>76°14’ N</td>
<td>119°20’ W</td>
<td>15</td>
</tr>
<tr>
<td>3</td>
<td>KTL</td>
<td>Kotelnny Island, Siberia</td>
<td>Russia</td>
<td>76°06’ N</td>
<td>137°36’ E</td>
<td>5</td>
</tr>
<tr>
<td>4</td>
<td>BRW</td>
<td>Point Barrow, Alaska</td>
<td>U.S.</td>
<td>71°19’ N</td>
<td>156°36’ W</td>
<td>11</td>
</tr>
<tr>
<td>5</td>
<td>STM</td>
<td>Ocean Station “M”</td>
<td>Norway</td>
<td>66°00’ N</td>
<td>2°00’ E</td>
<td>6</td>
</tr>
<tr>
<td>6</td>
<td>CBA</td>
<td>Cold Bay, Alaska</td>
<td>U.S.</td>
<td>55°12’ N</td>
<td>162°43’ W</td>
<td>25</td>
</tr>
<tr>
<td>7</td>
<td>SHM</td>
<td>Shemya Island</td>
<td>U.S.</td>
<td>52°43’ N</td>
<td>174°06’ E</td>
<td>40</td>
</tr>
</tbody>
</table>

\[
NMSD = \frac{1}{12} \sum_{m=1}^{12} \left( \frac{([\text{TBM}]_m + [\text{FOS}]_m + [\text{OCE}]_m - [\text{OBS}_m])^2}{\sigma_m^2} \right),
\]

where \([\text{TBM}]_m\), \([\text{FOS}]_m\), and \([\text{OCE}]_m\) are the monthly CO₂ concentrations resulting from the corresponding biospheric, fossil fuel, and ocean flux, respectively, \([\text{OBS}]_m\) is the mean observed value of CO₂ (from available data in the 1983 to 1992 period) and \(\sigma_m\) is the standard deviation of the observed value for each month \((m = 1, 2, \ldots, 12)\) of the year.

In this study we used the HAMOCC3 ocean biogeochemical model (Maier-Reimer 1993; Six & Maier-Reimer 1996) and the TM2 atmospheric transport model (Heimann 1995). The description and application of these models for purposes of simulating the seasonal cycle of atmospheric CO₂ is documented in Heimann et al. (1998). The three TBM's we applied in this study include the Carnegie-Ames-Stanford Approach (CASA; Randerson et al. 1997), Century (Parton et al. 1993), and the Terrestrial Ecosystem Model (TEM; McGuire et al. 1997; Tian et al. 1999). The data sets used as driving variables for each model are the same as used in the “Potsdam 95” comparison of NPP among global TBM's (Cramer et al. 1999), a comparison which included the three TBM's in this study. Differences in the absolute magnitude of global and regional NPP among the three TBM's have been described in Cramer et al. (1999) and Kicklighter et al. (1999). One important difference among the models concerns controls over the phenology of NPP. In CASA, the phenology of NPP is largely determined from satellite-derived data. In contrast, Century and TEM implement prognostic phenology algorithms. To facilitate evaluation of the seasonality of NPP and \(R_H\) among the models in this study, we focus our comparison on the relative proportion of annual NPP and annual \(R_H\) that is estimated to occur in each month of the year within
northern high and temperate latitudes. In the following sections we provide additional details on the description of the TBMs.

The Carnegie-Ames-Stanford Approach (CASA)

The CASA model calculates NPP from satellite-derived photosynthetically active radiation using a light use efficiency model (Field et al. 1995, 1998). For each monthly time step, a globally uniform maximum light use efficiency ($\epsilon^*$) is reduced by temperature and soil moisture stress scalars that reflect local environmental conditions. The calibration process involves adjusting $\epsilon^*$ to minimize the difference between observed and modeled NPP estimates across a number of sites worldwide. Simulated NPP in CASA is allocated to three living biomass pools, which include leaves, fine roots, and wood (Thompson et al. 1996).

Biomass is delivered to the soil as leaf litterfall, coarse woody debris, and fine root turnover (Randerson et al. 1996). Transformations of soil organic matter are represented following the structure of the Century model, with a suite of active and slow organic matter fractions (Parton et al. 1993). In the version of CASA used in this study, the ratio of carbon to nitrogen in leaf litterfall is held constant for each biome type (Randerson et al. 1996). A $Q_{10}$ of 1.5 is used to describe the response of microbial activity to surface air temperature (Raich & Potter 1995).

Century

The Century model uses information on climate, atmospheric CO$_2$, and nitrogen inputs to estimate the monthly fluxes and pools of carbon and nitrogen in terrestrial ecosystems (Parton et al. 1987, 1993). In this study, we applied version 4.0 of Century, which is described by Parton et al. (1993). In Century, maximum plant production is controlled by soil temperature, available water, and leaf area. A temperature-production function is specified according to plant functional types, such as C$_3$ cool season plants or C$_4$ warm season plants. Production is further modified by the current amount of above-ground plant material (i.e., self-shading), atmospheric CO$_2$ concentrations, and available soil nitrogen. To simulate savanna and shrubland ecosystems, grass and forest components compete for water, light, and nutrients in a prescribed manner. Simulated NPP in Century is allocated to eight vegetation pools.

Tissues that senescence from the vegetation pools of Century enter the soil as plant residues. The Century model simulates the decomposition of plant residues with a detailed submodel (13 pools) that divides soil organic carbon into three fractions: an active soil fraction (<10-year turnover time)
consisting of live microbes and microbial products; a protected fraction (decadal turnover time) that is more resistant to decomposition as a result of physical or chemical protection; and a fraction that has a very long turnover time (millennial turnover time). The decomposition of each soil organic fraction is calculated at a monthly time step as a function of the soil organic carbon in the fraction, air temperature, and soil moisture. A $Q_{10}$ of approximately 1.6 is used to describe the response of microbial activity to surface air temperature.

*The Terrestrial Ecosystem Model (TEM)*

The TEM is a highly aggregated TBM that uses spatially referenced information on climate, elevation, soils, and vegetation to make monthly estimates of important carbon and nitrogen fluxes and pool sizes in the global terrestrial biosphere. In this study we applied version 4.1 of TEM, in which the formulations for autotrophic and heterotrophic respiration have been modified from earlier versions as described in McGuire et al. (1997) and Tian et al. (1999). In TEM, NPP is the difference between carbon captured from the atmosphere as gross primary production (GPP) and carbon respired to the atmosphere by the vegetation. Monthly GPP is initially calculated in TEM as a function of photosynthetically active radiation, air temperature, atmospheric carbon dioxide concentration, and moisture availability. If nitrogen supply, which is the sum of nitrogen uptake and labile nitrogen in the vegetation, cannot meet the stoichiometric carbon to nitrogen ratio of biomass production, then GPP is reduced to meet the constraint. In the case where nitrogen supply does not limit biomass production, nitrogen uptake is reduced so that nitrogen supply meets the constraint of biomass production. In this way, the carbon-nitrogen status of the vegetation causes the model to allocate more effort toward either carbon or nitrogen uptake (McGuire et al. 1992). Plant respiration, which includes growth and maintenance respiration, is a function of GPP, vegetation carbon, and surface air temperature.

Biomass is delivered to the soil as litter production, which is calculated as a linear function of vegetation carbon (Raich et al. 1991). In TEM, the flux $R_H$ represents decomposition of all organic matter in an ecosystem and is calculated at a monthly time step as a function of one soil organic carbon compartment, air temperature, and soil moisture. Although the model does not track separate pools of soil organic matter, the formulations in the model implicitly consider issues of organic matter quality and turnover (McGuire et al. 1995, 1997). A $Q_{10}$ of 2.0 is used to describe the response of microbial activity to surface air temperature (Raich et al. 1991).
Results

For northern high latitudes, the baseline simulations of all three models estimate that monthly NPP is positive between May and September, with very low or negative values estimated during other months of the year (Figure 2(a–c)). The CASA baseline simulation estimates that approximately 40% of annual NPP in northern high latitudes occurs in July, with June and August as the next two most productive months. In contrast, both the Century and TEM baseline simulations estimate that maximum production occurs in June, with July and August as the next two most productive months. For all three models, implementation of the snowpack formulation has little effect on the proportion of annual NPP that is estimated to occur in each month.

In the baseline simulations of all three models, positive values of monthly $R_H$ are estimated for each month of the year for northern high latitudes, with the lowest monthly estimates in December and January and the highest estimates in July (Figure 2(d–f)). Although minimum and maximum $R_H$ of the snowpack simulations of the three models tends to occur in the same months as in the baseline simulations, the proportion of annual $R_H$ estimated by the snowpack simulations is lower between May and September and higher between November and March.

For northern temperate latitudes, the baseline simulation of all three models estimate positive values of NPP between February and November (Figure 3(a–c)). For December and January, the CASA and Century baseline simulations estimate low positive NPP while the TEM baseline simulation estimates low negative NPP. The baseline simulation of CASA estimates that July is the most productive month in northern temperate latitudes with June and August being the next two most productive months. In contrast, the baseline simulations of Century and TEM estimate that maximum production occurs in May and June, respectively. Similar to northern high latitudes, implementation of the snowpack formulation in each model has little effect on the proportion of annual NPP that is estimated to occur in each month.

Similar to northern high latitudes, positive values of monthly $R_H$ are estimated for each month of the year for northern temperate latitudes by the baseline simulation of each model, with the lowest monthly estimates in December and January and the highest estimates in July (Figure 3(d–f)). Although minimum and maximum $R_H$ of the snowpack simulations of the three models tends to occur in the same months as in the baseline simulations, the proportion of annual $R_H$ estimated by the snowpack simulations is lower between April and October and higher between November and March.

The differences in the monthly proportions of annual $R_H$ between the baseline and snowpack simulations cause differences in the seasonal uptake and release of $CO_2$ between the two simulations of all three models in
Figure 2. Relative proportions of annual net primary production (NPP) and heterotrophic respiration ($R_H$) that occur during each month of the year for baseline simulations (open bars) and snowpack simulations (hatched bars) between 60° N and 90° N for (a, d) the Carnegie-Ames-Stanford Approach (CASA), (b, e) Century, and (c, f) the Terrestrial Ecosystem Model (TEM).
Figure 3. Relative proportions of annual net primary production (NPP) and heterotrophic respiration \( (R_H) \) that occur during each month of the year for baseline simulations (open bars) and snowpack simulations (hatched bars) between 30° N and 60° N for (a, d) the Carnegie-Ames-Stanford Approach (CASA), (b, e) Century, and (c, f) the Terrestrial Ecosystem Model (TEM).
both northern high latitudes and northern temperate latitudes. For northern high latitudes, the snowpack simulations of all three models estimate greater uptake between June and August and greater release between November and March in comparison to the baseline simulations (Figure 4(a–c)). For comparison with Oechel et al. (1997), we analyzed the releases of CO₂ estimated for each of the models between October and May in northern high latitudes. Between October and May, the snowpack simulations estimate that the release of CO₂ from high latitude soils is 1241 \( 10^{12} \) g C (Tg C) for CASA, 543 Tg C for Century, and 1219 Tg C for TEM, which is 36%, 124%, and 41% higher than the estimates in the baseline simulations, respectively. Specifically in wet/moist tundra, the release of CO₂ estimated between October and May by the snowpack simulations is 112 g C m\(^{-2}\) for CASA, 50 g C m\(^{-2}\) for Century, and 62 g C m\(^{-2}\) for TEM. Similar to northern high latitudes, the snowpack simulations of all three models estimate greater uptake between May and August and greater release between November and March in northern temperate latitudes (Figure 4(c–e)). In comparison to the baseline simulations, the greater uptake of CO₂ from the atmosphere during the growing season and the greater release CO₂ to the atmosphere during the nongrowing season estimated by the snowpack simulations for terrestrial ecosystems north of 30° N has implications for the simulation of the seasonal cycle of atmospheric CO₂ at high latitude monitoring stations.

In comparison to the CO₂ concentrations simulated with the baseline fluxes of each biogeochemical model, the CO₂ concentrations simulated using the snowpack fluxes are generally in better agreement with observed concentrations between August and March at each of seven monitoring stations located in northern high latitudes (Figures 5, 6, and 7). There appears to be little effect of the snowpack simulations on the CO₂ concentrations simulated for May, June, and July. Among the seven high latitude stations, the normalized mean squared deviation between estimated and observed CO₂ concentrations was significantly improved for CO₂ fluxes associated with the snowpack simulation of each terrestrial biosphere model (Figure 8; for each TBM, Wilcoxin-signed rank test: \( W = -28.0 \), \( P = 0.016 \), \( N = 7 \)). Thus, representation of the insulative effects of snowpack in TBM stores generally improves simulation of atmospheric CO₂ concentrations in high latitudes during both the late growing season and the nongrowing season, and has little effect on concentrations at the beginning of the growing season.

**Discussion**

The observation that heterotrophic respiration during the nongrowing season represents a substantial proportion of annual heterotrophic respiration has
Figure 4. Monthly net ecosystem production (NEP) for baseline simulations (open bars) and snowpack simulations (hatched bars) from (a, b, c) 60°N to 90°N and (d, e, f) 30°N to 60°N for the Carnegie-Ames-Stanford Approach (CASA), Century, and the Terrestrial Ecosystem Model (TEM).
Figure 5. Comparison between the observed seasonal cycle of CO₂ and the simulated seasonal cycle produced by coupling the monthly estimates of net ecosystem production estimated by the Carnegie-Ames-Stanford Approach (CASA) and fossil fuel emissions with the Hamburg ocean and atmospheric transport models for each of the seven high latitude monitoring stations. The first six months of each cycle are displayed twice to reveal the annual variation more clearly. Mean and standard deviation are shown for the observed data. Station abbreviations are identified in Table 1.
Figure 6. Comparison between the observed seasonal cycle of CO₂ and the simulated seasonal cycle produced by coupling the monthly estimates of net ecosystem production estimated by the Century model and fossil fuel emissions with the Hamburg ocean and atmospheric transport models for each of the seven high latitude monitoring stations. The first six months of each cycle are displayed twice to reveal the annual variation more clearly. Mean and standard deviation are shown for the observed data. Station abbreviations are identified in Table 1.
Figure 7. Comparison between the observed seasonal cycle of CO₂ and the simulated seasonal cycle produced by coupling the monthly estimates of net ecosystem production estimated by the Terrestrial Ecosystem Model (TEM) and fossil fuel emissions with the Hamburg ocean and atmospheric transport models for each of the seven high latitude monitoring stations. The first six months of each cycle are displayed twice to reveal the annual variation more clearly. Mean and standard deviation are shown for the observed data. Station abbreviations are identified in Table 1.
Figure 8. Comparison of the normalized mean-squared deviation between the baseline and snowpack simulations of the seasonal cycle of atmospheric CO$_2$ at the seven high latitude monitoring stations for (a) the Carnegie-Ames-Stanford Approach (CASA), (b) Century, and (c) the Terrestrial Ecosystem Model (TEM). The normalized mean-squared deviation (nmsd) is an index that indicates the degree of difference between the observed and simulated monthly concentrations of atmospheric CO$_2$ at a monitoring station, i.e., a nmsd of zero indicates no difference.

been recognized for decades (Kelley et al. 1968; Coyne & Kelley 1974). More recently, investigations by Oechel et al. (1997), Jones et al. (1999), and Grogan and Chapin (pers. comm.) have attempted to quantify heterotrophic respiration of tundra ecosystems during the nongrowing season. Oechel et al. (1997) used chambers to measure CO$_2$ exchange in wet sedge and tussock tundra in October, November, January, and May. From these measurements, Oechel et al. (1997) estimated releases of 20 g C m$^{-2}$ from wet sedge tundra and 69 g C m$^{-2}$ from tussock tundra between October and May. These estimates were spatially extrapolated by Oechel et al. (1997) to calculate the global release of CO$_2$ for wet sedge tundra (19 Tg C) and tussock tundra (60 Tg C) between October and May. In contrast to Oechel et al. (1997), Jones et al. (1999) used an IRGA connected to a stainless steel probe to measure CO$_2$ concentrations and applied a diffusion model to calculate the CO$_2$ release. From these measurements, Jones et al. (1999) estimated releases of at least 9.4 g C m$^{-2}$ for coastal plain tundra and 8.4 g C m$^{-2}$ for upland tussock between October and May. Grogan and Chapin (pers. comm.), who used a technique based on soda lime, estimated winter releases of 120 to 190 g C m$^{-2}$ for three tussock tundra sites in Alaska. The releases of CO$_2$ from high latitude tundra between October and May estimated by the snowpack simulations of CASA (112 g C m$^{-2}$), Century (50 g C m$^{-2}$), and
TEM (62 g C m$^{-2}$) are within the range of measured CO$_2$ releases from tundra during the nongrowing season.

The variability among the estimates by Oechel et al. (1997), Jones et al. (1999), and Grogan and Chapin (pers. comm.) represents, in part, measurement error and spatial heterogeneity. Measurement error is one reason why the accuracy and precision of fluxes estimated by TBMs are difficult to evaluate with ground-based measurements. Spatial heterogeneity in fluxes across the landscape (Oechel et al. 1997; Jones et al. 1999) is another reason why additional variation in the observations will be unexplained by TBMs, which make estimates at much coarser spatial resolutions (0.5°) than the resolution of measurements (m$^2$ to hectares). Because of these two sources of variation, it is difficult for coarse-scale models to explain much more than 50% of the variation in observations that have not been used in model development or calibration (see Raich et al. 1991; Melillo et al. 1993; Kicklighter et al. 1994). Because of this limitation, it is difficult to evaluate the large-scale performance of TBMs with ground-based measurements, which are more effectively evaluated with site-specific modeling approaches (e.g., see Waelbroeck 1993; Waelbroeck & Louis 1995; Waelbroeck et al. 1997). Comparison of simulated CO$_2$ concentrations to observations at monitoring stations is one way to evaluate the large-scale seasonal dynamics of TBMs.

Although high latitude monitoring stations integrate seasonal CO$_2$ exchanges across much of the northern hemisphere, boreal forest and tundra tend to contribute more per unit area than temperate regions (see Kaminski et al. 1996). In addition, each of the high latitude stations is influenced by different regions within the northern hemisphere. For example, in comparison to other stations, the seasonal dynamics at Point Barrow are more enhanced by contributions from the Alaska peninsula and the dynamics at Shemya are influenced substantially by exchange in the eastern Asian continent (Kaminski et al. 1996). Because different high latitude stations integrate different regions of the globe, the simulation of atmospheric CO$_2$ provides a check on both the spatial and temporal qualities of CO$_2$ exchange by a TBM. The simulation of atmospheric CO$_2$ concentrations in this study indicates that the insulative effects of snowpack, as implemented in the TBMs in this study, improved the simulation of the seasonal dynamics of atmospheric CO$_2$ at high latitude monitoring stations during the late growing season and the nongrowing season.

Discrepancies between simulated and observed concentrations of atmospheric CO$_2$ in our simulations may, in part, be attributable to the ocean carbon cycle model and atmospheric transport model we used in this study. Because the seasonality of oceanic carbon exchange has little influence on northern hemisphere monitoring stations (Heimann et al. 1998), it is unlikely
that discrepancies were caused by the HAMOCC3. In contrast, simulated CO$_2$ concentrations at high latitude stations may be sensitive to the simulation of atmospheric transport. There are a number of approaches to modeling atmospheric transport, and a recent intercomparison among global transport models (Law et al. 1996) revealed substantial differences in the concentration fields computed from the same prescribed surface sources by different transport models. The TM2 model in many respects performed close to the “typical” coarse grid transport model (Law et al. 1996). Differences between simulated and observed concentrations may, in part, be attributed to the extraction of CO$_2$ concentrations from a coarse grid transport model. Although it would be useful to assess the sensitivity of seasonal cycle simulations to transport models that operate at finer spatial resolutions, it is beyond the scope of our study.

Because we use potential vegetation to describe the land surface, our analyses do not fully consider carbon release associated with land-use nor carbon uptake associate with aorestation, nitrogen deposition, CO$_2$ fertilization, or other factors (see Schimel 1995). Although there are noticeable effects of tropical vegetation fires at certain tropical stations (Jacobs et al. 1994; Wittenberg et al. 1998), there is little effect on simulations of the seasonal cycle at high latitude stations when fires in the boreal forest region are also considered (Wittenberg et al. 1998; McGuire, unpublished). Although net uptake of carbon in some regions is relevant to simulating the growth in the amplitude of atmospheric CO$_2$ that has been observed at some high latitude monitoring stations (Randerson et al. 1997), other issues are important to improving simulations of atmospheric CO$_2$ across all high latitude monitoring stations (Heimann et al. 1998).

A number of studies have used satellite-derived data from the Advanced Very High Resolution Radiometer (AVHRR) in modeling the seasonal signal of atmospheric CO$_2$ (Fung et al. 1987; Heimann & Keeling 1989; Knorr & Heimann 1995; Hunt et al. 1996; Randerson et al. 1996, 1997; Heimann et al. 1998; Sitch & McGuire, unpublished). In comparison to Century and TEM, the use of satellite-derived data is an important factor in the closer agreement between observed and estimated concentrations of atmospheric CO$_2$ for simulations using fluxes from CASA. Heimann et al. (1998) also observed that simulations using fluxes from the simple diagnostic biosphere model of Knorr and Heimann (1995), which uses satellite-derived data to define the phenology of NPP, fit the seasonal signal better in comparison to simulations that used fluxes from the five TBMs in the study that prognostically defined the phenology of NPP. In the BIOME3 model (Haxeltine & Prentice 1996), the use of satellite-derived phenology improves the timing of the beginning of the growing season to substantially improve simulation
of the seasonal dynamics of atmospheric CO₂ at high latitude monitoring stations in comparison with phenology that is prognostically defined (Sitch and McGuire, unpublished). Although the simulation of Sitch and McGuire (unpublished) with satellite-derived phenology substantially improves the timing of the beginning of the growing season, it only partially improved estimates of CO₂ concentrations during the growing season and had no effect on CO₂ concentration during the nongrowing season. Our results are complementary to the simulations of Sitch and McGuire (unpublished), i.e., at high latitude monitoring stations the implementation of the snowpack formulation had: (1) little effect on the timing of the beginning of the growing season; (2) improved estimates of CO₂ concentrations during the late growing season, and (3) substantially improved the estimates of CO₂ concentrations during the nongrowing season.

In this study we have demonstrated that processes controlling the release of CO₂ from soils during the nongrowing season are relevant in the context of the global carbon cycle. This finding complements site-specific investigations of this issue (Kelley et al. 1968; Coyne & Kelley 1974; Waelbroeck 1993; Waelbroeck & Louis 1995; Waelbroeck et al. 1997; Oechel et al. 1997; Jones et al. 1999; Grogan & Chapin, pers. comm.). Although the implementation of the insulative effects of snowpack in this study allow us to simulate CO₂ concentrations at high latitude monitoring stations with more accuracy, it is important to recognize that the simple algorithm is a first-order approximation of the effects of snowpack on CO₂ release from cold season soils. A number of mechanisms may contribute to the release of CO₂ to the atmosphere from cold season soils. These include microbial activity at low temperature (Mazur 1980; Coxson & Parkinson 1987; Zimov et al. 1993, 1996; Brooks et al. 1995, 1996, 1997; Schimel & Clein 1996; Oechel et al. 1997), CO₂ release during phase changes from water to ice in freezing soil (Coyne & Kelley 1971; Oechel et al. 1997), freeze-thaw processes (Schimel & Clein 1996; Goulden et al. 1998), and spring thaw (Coyne & Kelley 1974; Kling et al. 1991; Goulden et al. 1998). Although we know that each of these processes may contribute to cold season release of CO₂ from soils, the relative contribution of these different mechanisms is not well understood (Jones et al. 1999). Studies that elucidate the relative importance of mechanisms controlling the release of CO₂ from snow-covered soils should be coupled with full-year measurements of CO₂ exchange using eddy covariance towers. Recently, year-round measurements of CO₂ exchange with tower technology have been made in tundra (Oechel and Vourilhitis, unpublished) and boreal forest (Goulden et al. 1998). The capability to make year-round measurements of CO₂ exchange in snow-covered soils represents an important advance in the ability to develop and test the nongrowing season dynamics of
process-based models. The understanding gained from simultaneous tower and process studies is essential for identifying which processes should be incorporated in global TBMs.

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