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The observed increase in the seasonal amplitude of atmospheric CO₂ in the 3 northern latitudes reveals major ecological changes that are not captured by terrestrial 4 ecosystem models participating in the fifth phase of the Coupled Model Intercomparison 5 Project (CMIP5) (Graven et al., 2013). Here, we used atmospheric CO₂ records from 26 6 northern hemisphere stations with coverage longer than 15 years, and the LMDZ4 7 8 atmospheric transport model prescribed with Net Ecosystem CO₂ Exchange (NEE) from an ensemble of nine terrestrial ecosystem models, to attribute change in the seasonal 9 amplitude of atmospheric CO₂. We found significant (P<0.05) increases in seasonal 10 peak-to-trough CO₂ amplitude (AMP_{P-T}) at 9 stations, and in trough-to-peak amplitude 11 (AMP_{T-P}) at 8 stations over the last three decades. Most of the stations with an 12 increasing amplitude are located in Arctic and boreal regions (>50°N), consistent with 13 previous observations that the amplitude increased faster at Barrow (Arctic) than at 14 Mauna Loa (subtropics). The multi-model ensemble mean (MMEM) shows that the 15 physiological response of ecosystems to rising CO₂ concentration (eCO₂) and climate 16 change are dominant drivers of the increase in AMP_{P-T} and AMP_{T-P} in the high latitudes. 17 At the Barrow station, the observed increase of AMP_{P-T} and AMP_{T-P} over the last 33 18 years is explained by eCO₂ (39% and 42%) almost equally than by climate change (32% 19 and 35%). The increased carbon losses during carbon release period in response to 20 eCO₂ are related to an enhancement of ecosystem respiration due to the eCO₂ caused 21 increase in carbon storage during carbon uptake period. We also found smaller 22 contributions of air-sea CO₂ fluxes (10% for AMP_{P-T} and 11% for AMP_{T-P}) and 23 marginally significant impacts of change in land use (3% for AMP_{P-T} and 4% for 24 AMP_{T-P}) at Barrow, highlighting the important role of these factors in regulating change 25 in seasonal cycle of global carbon cycle, which has been generally ignored by previous 26 studies. 27

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1 Introduction

As an integrated signal of large scale ecological changes, the change in seasonal 2 variations of atmospheric CO₂ concentration is an emerging property of the carbon cycle (e.g. 3 Bacastow et al., 1985; Kohlmaier et al., 1989; Keeling et al., 1996; Randerson et al., 1997; 4 5 Piao et al., 2008; Graven et al., 2013; Zeng et al., 2014; Gray et al., 2014; Barlow et al., 2015; Forkel et al., 2016). The seasonal CO₂ amplitude (AMP) in the lower troposphere has 6 increased by $\approx 50\%$ north of 45°N since the 1960s (Graven et al., 2013), and this signal can 7 only be explained by an increased seasonality of net biome productivity (NBP) in boreal and 8 9 northern temperate ecosystems. Yet, there is no consensus on the major factors driving the increase of NBP and the quantitative contribution of other, smaller fluxes like fossil CO₂ 10 emissions and air-sea exchange. On the one hand, Gray et al., (2014) and Zeng et al., (2014) 11 suggested that agricultural improvements contributed to the increase of AMP by increasing 12 the seasonal NBP uptake in cultivated lands, but there is a two-fold difference in the 13 14 estimated contribution of this mechanismin in the two studies (range 17%-45% of the increasing AMP). On the other hand, Randerson et al. (1997) and Forkel et al. (2016) 15 showed that during the last three decades, most of the increase of amplitude took place at 16 stations north of 55°N. In this view, agriculture improvement seems unlikely to be the only 17 driving factor, because croplands are mainly in northern temperate latitudes (Foley et al., 18 2005). Using the LPJmL carbon cycle model with an improved phenology module coupled 19 with the TM3 atmospheric transport model, Forkel et al. (2016) found that it is mainly the 20 physiological response of northern plants to warming rather to increasing CO₂ that explains 21 the trend of AMP over the last 20 years, which is partly inconsistent with the observation 22 made by Graven et al. (2013) that AMP increased in the 1960s to the mid-1970s at a time 23 when northern temperature slightly decreased, highlighting the need to search deeper in the 24 attribution of the AMP trend. 25

In this paper, we investigate the AMP trend in the northern hemisphere over the last thirty years (1980-2012) using an ensemble of ecosystem models with different parameterizations of the effect of elevated CO₂ and climate change (TRENDYv2) (Sitch et al., 2015) with another transport model (LMDZ4) (Hourdin et al., 2013). We also separate the contribution of fossil fuel CO₂ emissions, air-sea fluxes as well as the effect of climate change, rising CO₂ (eCO₂), nitrogen deposition in some models and land use change on the
trends in the seasonality of land ecosystem exchange. The contribution of atmospheric
transport trends to AMP trends is further analyzed. Trends in seasonal atmospheric CO₂ from
26 northern (north of 23°N) long-term atmospheric stations (>15 years in the period of
1980-2012) of the NOAA-ERSL surface flask air-sampling network (Table S1 and Figure S1)
are used.

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8 **Results and Discussion**

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10 **Observed** CO₂ data

The 26 northern (north of 23°N) atmospheric stations selected are shown in Figure S1 11 and Table S1. According to the shape of detrended CO₂ seasonal cycle (Thoning et al., 1989, 12 see methods) (Figure S2), we divided the amplitude into peak-to-trough (AMP_{P-T}, defined as 13 14 the difference between the peak value and the trough value of CO₂ seasonal cycle in a year) and trough-to-peak (AMP_{T-P}, defined as the difference between the trough value of CO₂ 15 seasonal cycle in a year and the peak value of the cycle in the next year). As shown in Figure 16 1a, a significant (P<0.05) positive trend of AMP_{P-T} ranging from 0.05 to 0.15 ppm yr⁻¹ is 17 found at 9 stations, 8 of them being north of 50°N. The other stations do not show significant 18 positive AMP_{P-T} trends and 5 stations show negative trends (the latter being significant at 19 only one station UUM). The significant increase in AMP_{P-T} reflects mainly an increasing 20 CO₂ drawdown (defined by the monthly net CO₂ concentration change) in June and July 21 (Figure S3). 22

The trends in AMP_{T-P} reported in Table S1 are similar to those of AMP_{P-T} , logically expected because we remove a long-term mean trend of each CO₂ time series (Figure. 1b). In total, 7 out of 8 stations with a significant (P<0.05) increase of AMP_{T-P} during 1980-2012 are located north of 50°N. The months of September and October are those during which most of the negative trend of AMP_{T-P} occurs at those sites (Figure S3). Overall, no sties show significant positive trend in AMP_{T-P} during the study period.

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30 Terrestrial ecosystem output and simulation of CO₂ amplitude trends

1 The Net Ecosystem CO₂ Exchange (NEE) from Eight dynamic global vegetation models (DGVMs) from TRENDYv2 (Le Quéré et al., 2014; Sitch et al., 2015) (Table S2) was 2 prescribed to the atmospheric transport model (LMDZ4) (Hourdin et al., 2013, See Methods). 3 Time-varying monthly NEE from each model under simulation S3 (driven by CO₂, climate 4 change and land-cover change) (Sitch et al., 2015), fossil fuel and cement emissions (Andres 5 et al., 2011; Boden et al., 2016), and interannual air-sea fluxes (Buitenhuis et al., 2010) were 6 7 prescribed to the global LMDZ4 transport model (Hourdin et al., 2013) with variable winds over the last 33 years. This is the T1 simulation (see Methods and Table 4), from which the 8 9 modeled CO₂ concentration field was sampled at each station and analyzed for amplitude changes just like in the observed time series. 10

Most of T1 simulations results (except with the ISAM and JULES ecosystem models) 11 produce a significant increase in AMP_{P-T} at boreal (north of 50°N) stations (Figure 1a), 12 though there are differences among models. In comparison with the observed average trend 13 $(0.094 \pm 0.033 \text{ ppm yr}^{-1})$ of AMP_{P-T} at the 8 boreal stations with a significant increase in 14 AMP_{P-T}, three models show a larger AMP_{P-T} positive trend (CLM4.5: 0.105 ± 0.046 ppm yr⁻¹; 15 LPJ: 0.101 ± 0.053 ppm yr⁻¹; VISIT: 0.101 ± 0.059 ppm yr⁻¹). At the three boreal stations 16 17 with no significant trend in observed AMP_{P-T} (BAL, MHD, SHM in Figure 1a), the T1 simulation results also correctly reproduce no trend (Figure 1a) except for ORCHIDEE at 18 MHD and VISIT at SHM. 19

Similar to trends in AMP_{P-T}, statistically significant increasing trough-to-peak CO₂ amplitude (AMP_{T-P}) is found in the T1 simulation results (except again for ISAM and JULES), consistent with the observed trends. For the simulations with ISAM and JULES, temperate sites have more statistically significant increasing trends of AMP_{T-P} than boreal and Arctic sites.

Overall, unlike previous studies that have shown a systematic underestimation of AMP trend by ecosystem models, namely the CMIP5 models (Taylor et al., 2012) and the MsTMIP models (Huntzinger et al., 2013; Wei et al., 2014) at high northern latitudes (Graven et al., 2013; Thomas et al., 2016), we found both underestimation and overestimation of AMP trends from the TRENDYv2 models (Figure S4). This phenomenon can be related to different climate forcing (between CMIP5 and other ensembles), partly

different terrestrial ecosystem models, and the transport simulation with a different transport
model (LMDZ4 here instead of TM3 and ACTM in Graven et al. (2013) and TM3 in Thomas
et al. (2016)).

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Effects of different factors on the trend in AMP_{P-T}

In order to separate the contribution of different driving factors on the trend of AMP_{P-T}, 6 we performed transport simulations with Net Ecosystem CO₂ Exchange (NEE) changes 7 caused by different factors from factorial runs of the TRENDYv2 models, respectively with 8 variable CO₂ only (eCO₂), variable CO₂ and climate, and variable CO₂, climate and land 9 cover change (Table S4, see Methods). To further separate the contribution of atmospheric 10 transport trends to AMP trends, the LMDZ4 transport model was used with variable transport 11 fields (Dee et al., 2011) but constant NEE, air-sea CO₂ flux and fossil fuel and cement 12 emissions of the year 1979, so that trends in AMP from this simulation can be attributed to 13 14 transport trends only.

The impact of climate change on NEE affecting AMP_{P-T} trends estimated from 15 multi-model ensemble mean (MMEM) varies among stations (Figure 2a). We find a positive 16 trend of AMP induced by climate change at boreal atmospheric stations (8 of 11 stations 17 north of 50°N (Figure. 2a and S5b). On average, climate change caused an enhancement of 18 0.015±0.025 ppm yr⁻¹ in AMP_{P-T} over boreal region (north of 50 °N) (Figure 3a), which is 19 about 20% of the observed AMP_{P-T} trend. It has been suggested that warmer temperature is 20 21 associated with higher vegetation productivity of boreal ecosystems in spring through advancing vegetation green-up date (Keeling et al., 1996), although such positive effect may 22 saturate (Piao et al., 2014; Fu et al., 2015). As shown in Figure S6a, for the BRW site (71 °N), 23 the effect of climate change on AMP_{P-T} is positive mainly during May and June. 24

By contrast at the temperate sites (in the band of 23-50 °N), the effect of climate change on the AMP_{P-T} trends is mainly negative (10 of the 15 stations), althoguh the impact is not statistically significant (except TAP significant at P<0.05 and ASK marginally significant at P<0.1). On average, climate change is modeled to cause a decrease of AMP_{P-T} of -0.012±0.040 ppm yr⁻¹ at stations in the temperate band (Figure 3a). Analysis of NEE impacted by climate change (Trendy models S2 – S1 simulations) shows that climate change

alone caused a decrease in CO₂ uptake from April to August over western and central US, in
Eastern Europe, northeast China and Mongolia (Fig. S7b), associated with declining soil
moisture driven by rising temperature and decreasing precipitation in these regions (Sitch et al., 2015).

In the simulations of CO₂ with MMEM, eCO2 causes a statistically significant increase 5 in AMP_{P-T} at 10 of the 11 boreal stations (Figure 2a), and the magnitude of trend in AMP_{P-T} 6 driven by eCO2 $(0.036\pm0.005 \text{ ppm yr}^{-1})$ is about twice as large as that caused by climate 7 change (Figure 3a). This larger effect of eCO2 than climate change on AMP_{P-T} trends in 8 9 boreal zone is also present in simulations with individual ecosystem model NEE (Figure S5a and b). This result do not support previous findings by the study of Forkel et al. (2016) 10 which suggest a more prominent signal of climate change than eCO2 in the observed 11 increase in AMP_{P-T} in high northern latitudes. However, we agree that climate change rather 12 than eCO2 causes the latitudinal difference of trend in AMP_{P-T}. As shown in Figure 3a, we 13 14 found that the magnitude of eCO2 effect to increase the trend of AMP in temperate regions $(0.028\pm0.023 \text{ ppm yr}^{-1})$ is comparable to that in boreal regions, although less stations (9 of 15 15) show statistically significant effect (Figure 2a). It should be noted that four TRENDY 16 17 models (CLM4.5, ISAM, LPX and OCN) considered carbon-nitrogen interactions and nitrogen deposition, thus the signal of eCO₂ derived from these models includes the 18 interactive effect of nitrogen deposition. Another simulation with nitrogen deposition using 19 CLM4 model (Oleson et al., 2010; Mao et al., 2013) (see Methods) predicts that, however, 20 21 the effect of nitrogen deposition on AMP_{P-T} trend is not significant (P<0.05) at all sites (Figure S8a), while this result depends on individual model parameterizations (Galloway et 22 al., 2008). Further studies based on multiple models with C-N interactions are needed. 23

Both forest inventory data and model simulation show that afforestation and forest regrowth after abandonment of agriculture in northern ecosystems have an important role in the regional and global carbon balance (Pan et al., 2011; Houghton et al., 2015; FAO, 2015). As shown in Figure S7c, most TRENDYv2 DGVMs (except ISAM) predict that land use change enhance net carbon uptake during the period from April to August in Eastern Europe, China and central and eastern United States. Accordingly, a significant (P<0.05) or marginally significant (P<0.10) positive effect of LUC on the trend in AMP_{P-T} is predicted

1 across 6 boreal sites and 3 northern temperate sites (Figure 2a and S5c), although the magnitude of signal is generally smaller than the effect of eCO₂ and climate change. Overall, 2 the positive increase of AMP_{P-T} attributed to land use change is similar between boreal 3 region $(0.007\pm0.009 \text{ ppm yr}^{-1})$ and northern temperate region $(0.004\pm0.008 \text{ ppm yr}^{-1})$ (Figure 4 3a), suggesting that the latitudinal difference in observed AMP_{P-T} increase $(0.07\pm0.05 \text{ ppm})$ 5 yr⁻¹ in boreal zone and 0.01±0.05 ppm yr⁻¹ in temperate zone) has little linkage with land use 6 change. It should be noted that, however, there are still large uncertainties in estimating 7 land-use-change effect on AMP_{P-T} trend, primarily owing to processes of land use change 8 and management not considered in some Trendy models (e.g., wood harvest, shifting 9 cultivation, peat fires) (Table S3) and the lack of some critical processes (e.g., human 10 settlement, erosion/redeposition, woody encroachment) in all models (Houghton et al., 11 2012). 12

Over the past thirty years, global CO₂ emissions from fossil fuel consumption have 13 increased from 5.3 Pg C yr⁻¹ in 1980 to 9.7 Pg C yr⁻¹ in 2012 (Boden et al., 2016, Figure 14 S9a). However, the pattern of change is not spatially uniform in northern hemisphere. A 15 substantial significant increase in annual fossil fuel CO₂ emissions is found over northern 16 temperate region, whereas a significant decline is found in boreal region (Figure S9a). This 17 heterogeneity is also found in the period of April to August, during which AMP_{P-T} is 18 calculated for most northern temperate and boreal stations (Figure S9b). As a result, effect of 19 changes in fossil fuel carbon emissions on the trend in AMP_{P-T} is opposite between 20 21 temperate and boreal sites, although most sites show non-significant trend in AMP_{P-T} caused by trends of fossil CO₂ emissions. As shown in Figure 2a, a negative effect of fossil fuel 22 emissions on AMP_{P-T} trend is simulated at temperate sites (13 of 15 sites show negative 23 trend with 3 significant sites and 1 marginally significant site) (Figue 2a), and a positive 24 effect at most boreal sites (8 of 11 sites). The absolute value of the AMP_{P-T} trend associated 25 with fossil fuel emissions is generally larger at temperate (average of -0.013 ± 0.022 ppm yr⁻¹) 26 compared to boreal sites (average of 0.003 ± 0.007 ppm yr⁻¹) (Figure 3a). 27

A recent study (Horton et al., 2015) demonstrated robust trends in sub-seasonal atmospheric circulation patterns over mid-latitude regions during 1979-2013, particularly in summer and autumn. Such changes in the large-scale atmospheric circulation may exert an

1 effect on the trend of CO₂ amplitude. As shown in Figure 2a, although only two stations 2 (UUM and IZO) show statistically significant impact of transport trends on the trend of AMP_{P-T}, the magnitude of transport caused AMP_{P-T} trend is comparable or even larger than 3 the effect of climate change and eCO₂ on NEE at some atmospheric stations, particularly in 4 the temperate zone. For example, at UUM, the observed decrease in AMP_{P-T} is primary 5 caused by change in wind (Figure 2a). In terms of effects air-sea fluxes on the trend of 6 AMP_{P-T}, a weak contribution to AMP trends was simulated across most of sites except at 7 BRW (0.010 ppm vr⁻¹, P<0.05, 10% of the observed trend) and MBC (0.015 ppm vr⁻¹, P<0.1, 8 9 16% of the observed trend).

The mechanisms driving the trend in AMP_{P-T} are here analyzed with observations at the 10 Arctic station of BRW (71°N), the longest northern high latitude CO2 record showing an 11 increase of amplitude of 35% since 50 years, larger than at the Mauna Loa longest record 12 located in the sub-tropics (e.g. Graven et al., 2013; Zeng et al., 2014; Gray et al., 2014; 13 Barlow et al., 2015; Forkel et al., 2016). Our transport simulations with multi-model 14 ensemble mean (MMEM) NEE produce that AMP_{P-T} at the BRW site significantly inreased 15 by about 0.095 ppm yr⁻¹ from 1980 to 2012, which is comparable with the observed trend of 16 0.097 ppm yr⁻¹ (Figure 1a). eCO₂ is identified as the largest contributor of increasing AMP_{P-T} 17 with a trend of 0.039 ppm yr⁻¹ (40% of the observed trend, P<0.05), followed by climate 18 change with a trend of 0.031 ppm yr⁻¹ (32% of the observed trend, P<0.05) (Figure S5a and 19 b). The effect of ocean flux is of 0.010 ppm yr⁻¹ (10% of observed trend, P<0.05), and land 20 use change has marginally significant contributions (0.003 ppm yr⁻¹ and 3% of observed 21 trend, P<0.1) (Figure S5c and e). For other factors such as fossil fuel emissions and transport 22 (Figure S5d and f), non-significant impacts on AMP_{P-T} trend were produced. 23

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25 Effects of different factors on trend in AMP_{T-P}

We also assessed the effect of different factors on trend in AMP_{T-P} with the same NEE and transport model simulation results (See Methods). In contrast to carbon uptake period, climate change accelerates carbon release from boreal ecosystems during the non-carbon uptake period. As shown in Figure 2b, an increasing AMP_{T-P} (a negative trend in AMP_{T-P} indicates a larger release) is simulated at 8 of the 11 boreal sites (1 site significant at P<0.05;

1 2 sites marginally significant at $0.05 \le P \le 0.1$). In contrast, a decreasing AMP_{T-P} (shown with positive trend) is produced at 12 of 15 temperate sites (1 site significant at P < 0.05; 1 site 2 marginally significant at 0.05<P <0.1) (Figure 2b). It has been suggested that autumn 3 warming enhance vegetation productivity through delaying vegetation senescence, as well as 4 accelerate ecosystem respiration (Piao et al., 2008; Vesala et al., 2010). Therefore, the 5 opposite effect of climate change on the trend of AMP_{T-P} in boreal region (-0.016±0.027 6 ppm yr⁻¹) and temperate region $(0.011\pm0.040 \text{ ppm yr}^{-1})$ (Figure 3b) is probably due to 7 different magnitude of the response of vegetation productivity (GPP) and ecosystem 8 9 respiration (TER) to climate change. Indeed, the model results show that the rising temperature induced increase of TER is greater than that of GPP in high northern latitudes, 10 whereas the increase of GPP is larger in temperate regions (Figure S10). 11

Atmospheric CO₂ simulations from MMEM NEE produce an increasing AMP_{T-P} in 12 response to eCO₂ at 25 of 26 temperate and boreal sites (19 sites significant at P<0.05; two 13 14 sites marginally significant at 0.05<P<0.1, Figure 2b). NEE from 6 out of 8 terrestrial ecosystem models (except ISAM and JULES) also produces an enhancing AMP_{T-P} from 15 eCO2 (Figure S11a). This result indicates an acceleration of carbon release during the 16 non-carbon uptake period as an indirect effect of the NEE response to eCO2. This is due to 17 the increment in carbon storage caused by the enhancement of net carbon uptake during the 18 carbon uptake period under the effect of eCO₂, which stimulates ecosystem respiration 19 during the non-carbon uptake period (Figure S12). Similar to the contribution of land use 20 21 change to the trends of AMP_{P-T}, we also found statistically significant enlargement of AMP_{T-P} in response to land use change at 9 of 26 sites (Figure 2b). 22

Similar to the effect on AMP_{P-T} , the contribution of fossil fuel CO_2 emissions, air-sea fluxes and transport on the trend of AMP_{T-P} is only statistically significant at a minority of sites (only 1, 4 and 2 sites significant at P<0.05 for the effect of fossil fuel, air-sea fluxes and transport, respectively) (Figure 2b). However, the magnitude of signal induced by transport and fossil fuel emissions is generally remarkable over temperate region (Figure 3b), causing an average impact of -0.014±0.036 ppm yr⁻¹ and 0.010±0.014 ppm yr⁻¹ in the trend of AMP_{P-T} , respectively.

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Overall, the observed significant enlargement of AMP_{T-P} at the BRW site (-0.090 ppm

yr⁻¹) is mainly driven by eCO2 (-0.038 ppm yr⁻¹ and 42% of observing trend, P<0.05),
climate change (0.032 ppm yr⁻¹ and 35% of observing trend, P<0.05), ocean flux change
(-0.010 ppm yr⁻¹ and 11% of observing trend, P<0.05) and land use change (-0.003 ppm yr⁻¹
and 4% of observing trend, P<0.05).

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6 Conclusion

7 It is difficult to attribute trends in seasonal atmospheric CO₂ concentration both because of regionally different contributions, and because terrestrial ecosystem models lack 8 9 mechanisms or did not report separately disturbance contributions and nitrogen deposition contributions. Unlike previous studies based on one model only (Zeng et al., 2014; Forkel et 10 al., 2016), our results based on an ensemble of models to capture the amplitude trends suggest 11 that rising atmospheric CO₂ concentration is the primary driver of enhancement of both 12 AMP_{P-T} and AMP_{T-P}, although climate change plays a critical role and contributes largely to 13 14 the latitudinal difference of AMP trend. In addition, the effect of other factors such as land use change, fossil fuel emissions, ocean flux, and transport on the trend in AMP_{P-T} and 15 AMP_{T-P} is not statistically significant at most of stations, but still large enough to cancel out 16 the effect of eCO_2 at some temperate sites where the observing seasonal CO_2 trends are small. 17 However, the large uncertainties in the forcing data on land use change and fossil fuel 18 emission at the moment do not allow an unequivocal statement on the contribution of these 19 factors, and further studies based on spatially and temporally explicit historic data sets 20 21 including land use and fossil fuel emission are needed. Finally, we found that rising atmospheric CO_2 concentration has opposite implication in the northern ecosystem carbon 22 balance between carbon uptake period (trend of AMP_{P-T}) and non-carbon uptake period (trend 23 of AMP_{P-T}), due to the lagged effects of increase in carbon storage during carbon uptake 24 period on carbon cycle in non-carbon uptake period. Our results not only provide insights for 25 large-scale field experiments, but also highlight the importance of understanding the carbon 26 releasing processes during the non-growing season, which is critical for reliable projection of 27 global carbon cycle, and thus, the future climate change. 28

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30 Methods

1 Atmospheric CO₂ concentration data. Weekly atmospheric CO₂ concentration data were obtained from the Earth System Research Laboratory, National Oceanic and 2 Atmospheric Administration (NOAA-ESRL) archive (Masarie et al., 2014) for the period of 3 1980-2012. 26 northern temperate and boreal stations with observations longer than 15 years 4 (Table S1) are included in our analyses, given the focus of our study is the long-term trend, 5 which would not be robust without long-term observations. The seasonal curves of 6 atmospheric CO₂ at each station were extracted by fitting the observation data with a function 7 consisting of quadratic polynomial for the long-term trend, four-harmonics for the annual 8 9 cycle, and a 80-days Full-Width Half-Maximum value (FWHM) averaging filter and a 390-days FWHM averaging filter to further remove short term variations and remaining 10 annual cycles still present in the residuals after the function fit (Thoning et al., 1989). The 11 processing process is incorporated in the standard CO₂ data processing software (CCGCRV) 12 developed by NOAA-ESRL (Thoning et al., 1989). From the seasonal curve of atmospheric 13 14 CO₂, we then obtained the amplitude and monthly concentration difference.

Land-atmosphere CO₂ exchange. An ensemble of eight dynamic global vegetation 15 models (DGVMs) from TRENDYv2 was used to simulate monthly land-atmosphere CO2 16 exchange (Net Biome Productivity, NBP) for the period 1979-2012. These models were 17 coordinated to perform three simulations (S1, S2 and S3) following the TRENDYv2 protocol 18 (Le Quéré et al., 2013; Sitch et al., 2015). In simulation S1, only atmospheric CO₂ was varied. 19 In simulation S2, only atmospheric CO₂ and climate were varied. In simulation S3, 20 21 atmospheric CO₂, climate and land use were varied. The effects of rising atmospheric CO₂, climate change and land use change on NBP can then be obtained from S1, the difference 22 between S2 and S1, and the difference between S3 and S2, respectively. Among the eight 23 TRENDY models, four models (CLM4.5, ISAM, LPX and OCN) considered carbon-nitrogen 24 interactions and nitrogen deposition through simulation S1 to S3. All models used the same 25 forcing datasets, of which global atmospheric CO₂ concentration was from the combination of 26 ice core records and atmospheric observations (Keeling & Whorf, 2005 and update); historical 27 climate fields were from CRU-NCEP dataset (http://dods.extra.cea.fr/data/p529viov/cruncep/); 28 land use data was from the Hyde database (Hurtt et al., 2011). The effect of nitrogen 29 deposition was derived from an additional simulation (S4) performed by CLM4 model 30

(Oleson *et al.*, 2010; Mao *et al.*, 2013) in which all the driving factors (atmospheric CO₂,
 climate and land use) were kept constant at 1980 value except transient nitrogen deposition
 from 1980 through 2012 (Lamarque *et al.*, 2005). Detailed information of the nine DGVMs
 used in this study is listed in Table S2.

Ocean-atmosphere CO₂ exchange. A biogeochemical model PlankTOM5 combined 5 with a global ocean general circulation model NEMO (NEMO-PlankTOM5) were used to 6 7 simulate the physical, chemical and biological processes that affect the surface ocean CO₂ concentration and thus the ocean-atmosphere CO₂ exchange (Buitenhuis et al., 2010; Le 8 9 Quéré et al., 2015). The PlankTOM5 model was forced by inputs of ions and compounds from river, sediment and dust (Cotrim da Cunha et al., 2007; Aumont et al., 2003). The 10 NEMO model was driven by daily wind and precipitation from NCEP reanalysis (Kalnay et 11 al., 1996). Further details can be found in Buitenhuis et al. (2010). 12

Fossil fuel CO₂ emission. A gridded monthly time series of fossil fuel CO₂ emissions 13 14 from CDIAC were constructed based on proportional-proxy approach (Andres et al., 2011; Boden et al., 2016). Firstly, available monthly fossil fuel consumption data from 21 countries 15 was compiled, which account for about 80% of global total emissions. Then these data were 16 17 used as a proxy for all remaining countries without monthly data based on countries' similarities in climates and economies (for few countries, geographic closeness was also 18 considered). For some years without explicit monthly data, Monte Carlo methods were used 19 to apply data in years with known monthly fractions to those missing-data years. Further 20 21 details can be found in Andres et al. (2011).

The atmospheric transport model. We used LMDZ4, a global tracer transport model (Hourdin et al., 2013) driven by the re-analysis 3-D atmospheric wind fields from the European Centre for Medium-Range Weather Forecasts (Dee et al., 2011), to transform land-atmosphere CO₂ exchange, fossil fuel CO₂ emission and ocean-atmosphere CO₂ exchange into point estimates of CO₂ concentration at 26 sites. The model configuration we used has a horizontal spatial resolution of 3.75° longitude × 2.5° latitude with 19 vertical layers.

To separate the effects of changes in atmospheric CO₂ ('CO₂'), climate ('CLIM'), land use ('LU'), fossil fuel ('FF'), ocean carbon flux ('Ocean') and atmospheric transport ('Wind')

on seasonal atmospheric CO₂ concentration change, we designed eight transport simulations 1 (T1~T8, see Table S3). The first one (T1) used time-varying monthly land-atmosphere CO₂ 2 exchange under scenario S3 (driven by rising CO₂, climate change and land use change), 3 fossil fuel CO₂ emission, and ocean-atmosphere CO₂ exchange coupled with LMDZ4 4 transport model having variable winds, indicating the combined effects of 'CO₂', 'CLIM', 5 'LU', 'FF', 'Ocean' and 'Wind'. To assess the contribution of 'Wind', the LMDZ4 transport 6 7 experiment was force by historical varying wind but constant land-atmosphere CO₂ exchange, fossil fuel CO₂ emission and ocean-atmosphere CO₂ exchange in 1979 (T6). Next, to 8 investigate the single effect of 'CO₂', 'CLIM' and 'LU', we utilized LMDZ4 model with 9 varying winds to perform another three transport simulations (T2, T3 and T4, see Table 3) in 10 which fossil fuel CO₂ emission and ocean-atmosphere CO₂ exchange were constant at 1979 11 value but land-atmosphere CO₂ exchange was varying under three scenarios (S1, driven by 12 CO₂; S2, driven by CO₂ and CLIM; S3, driven by CO₂, CLIM and LU). Consequently, the 13 single effect of 'CO₂' on seasonal CO₂ variation can be assessed by the difference between T2 14 and T6, that of 'CLIM' from the difference between T3 and T2, and that of 'LU' from the 15 difference between T4 and T3. In addition, we prescribed varying land-atmosphere CO₂ 16 17 exchange from CLM4 model under scenario S4 (only nitrogen deposition varying), constant fossil fuel CO₂ emission and ocean-atmosphere CO₂ exchange to LMDZ4 model with 18 constant winds (transport simulation T5) to obtain the effect of nitrogen deposition. Finally, to 19 gain the single effect of 'FF' and 'Ocean' on CO₂ seasonal variation, we performed another 20 21 two simulations in which only fossil fuel CO₂ emission or ocean-atmosphere CO₂ exchange was varying in addition to variable winds (T7 and T8). Thus the contribution of 'FF' can be 22 calculated from the difference between T7 and T6, and that of 'Ocean' from the difference 23 between T8 and T6. 24

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Figure 1 Observed and modeled trends in CO₂ seasonal peak-to-trough amplitude (AMP_{P-T}) (a) and trough-to-peak amplitude (AMP_{T-P}) (b). Here we calculated the modeled AMP_{P-T}/ AMP_{T-P} trends based on eight TRENDY models and multi-model ensemble mean (MMEM) under T1 transport simulation (see methods). Shown on the top of the figure are the abbreviated names of 26 atmospheric CO₂ concentration measurement sites in northern temperate and boreal regions. We sort the sites according to their latitudes, from 23°N to 90°N. Each row represents trends for different sites, while each column represents trends derived from observation and model simulations at a certain site. Gray grids show insignificant trends (P > 0.10), while colored grids without slashes indicate statistically significant (P < 0.05) and those with slashes marginal significant (P < 0.10). The number in each grid shows the value of the trend. Site abbreviations are defined in Table S1.

a	←						←23°N-50°N >														— 50°N-90°N ———					
	Stat	ter ter	MID	10	WIS .	BHNN	BINE	MLC	1 AS	p2P	JIP .	HUNR -	مرر	CMO	AUH	SHIM	MHD	CBA	BAL	(CE-	STM	BRW	SUM	MBC	1 ^{EP}	Ŕ
Observation	0.022	0.01	0.06	0.014	0.035	-0.002	0.088	0.015	0.008	0.014	0.047	-0.01	-0.145	-0.036	0.011	0.049	0.025	0.071	-0.025	0.055	0.055	0.097	0.148	0.097	0.128	0.1
CLM4.5	0.058	0.04	0.045	0.076	0.09	0.006	0.086	0.073	0.048	0.022	0.089	0.029	-0.062	0.093	0.185	0.034	0.021	0.095	0.188	0.153	0.085	0.117	0.061	0 189	0.062	0.07
ISAM	0.007	0.028	0,022	0.043	0.062	0.019	0.08	0.021	-0.029	-0.01	0.071	-0.009	-0.043	0.067	0.093	0.026	0.016	0.022	0.019	0.028	-0.048	0.005	-0.018	0.103	-0.032	-0.01
JULES	-0.06	0,039	0.021	0.058	0.091	-0.001	0.083	0.069	-0.138	-0.011	0.079	-0.009	-0,146	0.109	0.144	-0.001	-0.117	0.032	-0.198	0.118	-0.029	0.067	0.015	0.279	0.016	0.00
LPJ	-0.077	0.024	0.009	0.018	0.029	-0.055	0.08	0.219	-0.02	0.049	0.046	0.02	-0.122	0.041	-0.018	-0.063	-0.032	0.126	-0.083	0.107	0.042	0.183	0.02	0.148	0.095	0.0
LPX	0.021	0.01	0.004	0.022	-0.068	-0.024	-0.019	0.046	-0.059	0,028	0.058	0.006	-0.254	0.015	-0.016	-0.004	-0.014	0.035	0.052	0.037	0.04	0.065	0	0.114	0.019	0.03
OCN	-0.021	0.008	0.011	0.021	0.04	-0.034	-0.007	0.025	0.077	0,032	0.074	0.01	-0.153	0.082	0.285	0.025	0.031	0.042	0.052	0.08	0.043	0.063	0.038	0.089	0.072	0.03
ORCHIDEE	-0.013	0.01	-0.008	0.062	-0.111	-0.049	-0.035	0.011	0.082	0.085	0.094	0.038	-0.346	-0.121	0.279	-0.076	0.226	0.03	0.267	0.158	0.073	0.159	0.028	-0.085	0.24	0.11
VISIT	0.096	0.052	0.052	0.073	0.082	-0.018	0.096	0.162	0.072	0.03	0.091	0.036	-0.048	0.138	0.125	0.087	-0.019	0.1	0.169	0 111	0.022	0.099	0.044	0.164	0.161	0.04
MMEM	0.002	0.026	0.02	0.047	0.027	-0.02	0.045	0.018	0.004	0.028	0.075	0.015	-0.147	0.053	0.135	0.003	0.014	0.06	0.058	0.106	0.029	0.095	0.024	0.125	0.079	0.04
L.								•NI 50	0												50°N	00%				
D	ح	4	~	~	6	4.	—23 .4.	п-50 Со	-N	a		æ	10	.0			.0	6	.		50°N	-90-IN - 0	1.	0	0	, ,
	ASI	t,	MIL	10	2412	BUN	BUNT	Al.	1Pr	P	SIL	42	JUL	CNIC	AN.	SHIL	WHIL	CS,	BUT	Qr.	SIN	\$F	SUL	MBC	16	P
Observation	-0.027	-0.012	-0.062	-0.018	-0.026	-0.022	-0.108	0.003	-0.013	-0.003	-0.07	0.006	0.079	0.045	-0.055	-0.036	-0.042	-0.069	0.063	-0.053	-0.056	-0.09	-0.16	-0.141	-0.13	-0.09
CI M4 5	-0.066	-0.041	-0.044	-0.077	-0.091	-0.035	-0.111	-0.087	-0.124	-0.028	-0.075	-0.036	0.024	-0.047	-0.296	-0.049	0.03	-0.091	-0.2	-0.177	-0.076	-0.104	-0.058	-0.22	-0.052	-0.06
OLIMINO			77	//	0.005						//													-0.123	0.028	0.02
ISAM	-0.007	-0.031	-0.02	-0.044	-0.065	-0.021	-0.09	-0.02	-0.028	0.012	-0.063	0.009	0.008	-0.045	-0.003	-0.037	-0.022	-0.014	-0.026	-0.054	0.057	0.001	0.033			
ISAM JULES	-0.007 0.068	-0.031 -0.045	-0.02 -0.021	-0.044 -0.062	-0.065	-0.021 -0.002	-0.09 -0.087	-0.02 -0.045	-0.028 0.102	0.012 0.015	-0.063 -0.06	0.009 0.01	0.008	-0.045 -0.116	-0.003 0.092	-0.037 -0.027	-0.022 0.206	-0.014 -0.017	-0.026 0.208	-0.054	0.057 0.016	0.001	0.033	-0.326	-0.017	-0.00
ISAM JULES LPJ	-0.007 0.068 0.073	-0.031 -0.045 -0.027	-0.02 -0.021 -0.008	-0.044 -0.062 -0.028	-0.065 -0.108 -0.012	-0.021 -0.002 0.021	-0.09 -0.087 -0.104	-0.02 -0.045 -0.229	-0.028 0.102 -0.061	0.012 0.015 -0.059	-0.063 -0.06 -0.081	0.009 0.01 -0.028	0.008 0.136 0.127	-0.045 -0.116 -0.041	-0.003 0.092 0.002	-0.037 -0.027 0.03	-0.022 0.206 0.076	-0.014 -0.017 -0.133	-0.026 0.208 0.063	-0.054 -0.163 -0.153	0.057 0.016 -0.041	0.001 -0.077 -0.18	0.033 0 -0.019	-0.326 -0.227	-0.017 -0.096	-0.00
ISAM JULES LPJ LPX	-0.007 0.068 0.073 -0.027	-0.031 -0.045 -0.027 -0.009	-0.02 -0.021 -0.008 -0.007	-0.044 -0.062 -0.028 -0.019	-0.065 -0.108 -0.012 0.073	-0.021 -0.002 0.021 -0.005	-0.09 -0.087 -0.104 0.024	-0.02 -0.045 -0.229 -0.043	-0.028 0.102 -0.061 -0.009	0.012 0.015 -0.059 -0.035	-0.063 -0.06 -0.081 -0.053	0.009 0.01 -0.028 -0.011	0.008 0.136 0.127 0.217	-0.045 -0.116 -0.041 0.02	-0.003 0.092 0.002 -0.209	-0.037 -0.027 0.03 -0.015	-0.022 0.206 0.076 0.046	-0.014 -0.017 -0.133 -0.038	-0.026 0.208 0.063 -0.051	-0.054 -0.163 -0.153 -0.055	0.057 0.016 -0.041 -0.04	0.001 -0.077 -0.18 -0.066	0.033 0 -0.019 -0.007	-0.326 -0.227 -0.139	-0.017 -0.096 -0.03	-0.00 -0.0
ISAM JULES LPJ LPX OCN	-0.007 0.068 0.073 -0.027 0.004	-0.031 -0.045 -0.027 -0.009 -0.014	-0.02 -0.021 -0.008 -0.007 -0.011	-0.044 -0.062 -0.028 -0.019 -0.026	-0.065 -0.108 -0.012 0.073 -0.077	-0.021 -0.002 0.021 -0.005 0.006	-0.09 -0.087 -0.104 0.024 0.005	-0.02 -0.045 -0.229 -0.043 0.002	-0.028 0.102 -0.061 -0.009 -0.119	0.012 0.015 -0.059 -0.035 -0.036	-0.063 -0.06 -0.081 -0.053 -0.067	0.009 0.01 -0.028 -0.011 -0.018	0.008 0.136 0.127 0.217 0.095	-0.045 -0.116 -0.041 0.02 -0.071	-0.003 0.092 0.002 -0.209 -0.313	-0.037 -0.027 0.03 -0.015 -0.041	-0.022 0.206 0.076 0.046 -0.025	-0.014 -0.017 -0.133 -0.038 -0.044	-0.026 0.208 0.063 -0.051 -0.068	-0.054 -0.163 -0.153 -0.055 -0.095	0.057 0.016 -0.041 -0.04 -0.846	0.001 -0.077 -0.18 -0.066 -0.061	0.033 0 -0.019 -0.007 -0.05	-0.326 -0.227 -0.139 -0.102	-0.017 -0.096 -0.03 -0.082	-0.00 -0.09 -0.09
ISAM JULES LPJ LPX OCN ORCHIDEE	-0.007 0.068 0.073 -0.027 0.004 0.02	-0.031 -0.045 -0.027 -0.009 -0.014 -0.01	-0.021 -0.008 -0.007 -0.011 0.013	-0.062 -0.028 -0.019 -0.026 -0.063	-0.065 -0.108 -0.012 0.073 -0.077 0.084	-0.021 -0.002 0.021 -0.005 0.006 0.007	-0.09 -0.887 -0.104 0.024 0.005 0.026	-0.02 -0.045 -0.229 -0.043 0.002 0.003	-0.028 0.102 -0.061 -0.009 -0.119 -0.11	0.012 0.015 -0.059 -0.035 -0.036 -0.101	-0.063 -0.06 -0.081 -0.053 -0.067 -0.104	0.009 0.01 -0.028 -0.011 -0.018 -0.846	0.008 0.136 0.127 0.217 0.095 0.186	-0.045 -0.116 -0.041 0.02 -0.071 0.171	-0.003 0.092 0.002 -0.209 -0.313 -0.239	-0.037 -0.027 0.03 -0.015 -0.041 0.037	-0.022 0.206 0.076 0.046 -0.025	-0.014 -0.017 -0.133 -0.038 -0.044 -0.06	-0.026 0.208 0.063 -0.051 -0.068 -0.323	-0.054 -0.163 -0.153 -0.055 -0.095 -0.095	0.057 0.016 -0.041 -0.04 -0.846 -0.889	0.001 -0.077 -0.18 -0.066 -0.061 -0.163	0.033 0 -0.019 -0.007 -0.05 -0.032	-0.326 -0.227 -0.139 -0.102 0.098	-0.017 -0.096 -0.03 -0.082 -0.288	-0.00 -0.09 -0.09 -0.09 -0.09
ISAM JULES LPJ LPX OCN ORCHIDEE VISIT	-0.007 0.068 0.073 -0.027 0.004 0.02 -0.116	-0.031 -0.045 -0.027 -0.009 -0.014 -0.01	-0.021 -0.008 -0.007 -0.011 0.013 -0.048	-0.0844 -0.062 -0.028 -0.019 -0.026 -0.063 -0.073	-0.065 -0.108 -0.012 0.073 -0.077 0.084 -0.042	-0.021 -0.002 0.021 -0.005 0.006 0.007 -0.007	-0.09 -0.087 -0.104 0.024 0.005 0.026 -0.119	-0.02 -0.045 -0.229 -0.043 0.002 0.003 -0.199	-0.028 0.102 -0.061 -0.009 -0.119 -0.111 -0.18	0.012 0.015 -0.059 -0.035 -0.036 -0.101 -0.842	-0.063 -0.06 -0.081 -0.053 -0.067 -0.104 -0.077	0.009 0.01 -0.028 -0.011 -0.018 -0.846 -0.043	0.008 0.136 0.127 0.217 0.095 0.186 0.022	-0.045 -0.116 -0.041 0.02 -0.071 0.171 -0.117	-0.003 0.092 0.002 -0.209 -0.313 -0.239	-0.037 -0.027 0.03 -0.015 -0.041 0.037 -0.128	-0.022 0.206 0.076 0.046 -0.025 -0.291 0.112	-0.014 -0.017 -0.133 -0.038 -0.044 -0.06 -0.102	-0.026 0.208 0.063 -0.051 -0.068 -0.323 -0.18	-0.054 -0.163 -0.153 -0.055 -0.095 -0.172 -0.235	0.057 0.016 -0.041 -0.04 -0.044 -0.049 -0.014	0.001 -0.077 -0.18 -0.066 -0.061 -0.163 -0.892	0.033 0 -0.019 -0.007 -0.05 -0.032 -0.052	-0.326 -0.227 -0.139 -0.102 0.098 -0.17	-0.017 -0.096 -0.03 -0.082 -0.288 -0.156	-0.00 -0.03 -0.03 -0.03 -0.12 -0.04
ISAM JULES LPJ LPX OCN ORCHIDEE VISIT MMEM	-0.007 0.068 0.073 -0.027 0.004 0.02 -0.116 -0.006	-0.031 -0.045 -0.027 -0.009 -0.014 -0.014 -0.054 -0.029	-0.02 -0.021 -0.008 -0.007 -0.011 0.013 -0.048 -0.018	-0.044 -0.062 -0.028 -0.019 -0.026 -0.063 -0.073 -0.049	-0.065 -0.108 -0.012 0.073 -0.077 0.084 -0.042 -0.03	-0.021 -0.002 0.021 -0.005 0.006 0.007 -0.007 -0.004	-0.09 -0.887 -0.104 0.024 0.005 0.026 -0.119 -0.057	-0.02 -0.045 -0.229 -0.043 0.002 0.003 -0.199 -0.877	-0.028 0.102 -0.061 -0.009 -0.119 -0.111 -0.18 -0.066	0.012 0.015 -0.059 -0.035 -0.036 -0.101 -0.842 -0.834	-0.063 -0.081 -0.053 -0.067 -0.104 -0.077 -0.072	0.009 0.01 -0.028 -0.011 -0.018 -0.046 -0.043 -0.02	0.008 0.136 0.127 0.217 0.095 0.186 0.022 0.102	-0.045 -0.116 -0.041 0.02 -0.071 0.171 -0.117 -0.031	-0.003 0.092 0.002 -0.209 -0.313 -0.239 -0.254 -0.153	-0.037 -0.027 0.03 -0.015 -0.041 0.037 -0.128 -0.029	-0.022 0.206 0.076 0.046 -0.025 0.291 0.112 0.017	-0.014 -0.017 -0.133 -0.038 -0.044 -0.06 -0.102	-0.026 0.208 0.063 -0.051 -0.068 -0.323 -0.18 -0.072	-0.054 -0.163 -0.153 -0.055 -0.095 -0.095 -0.172 -0.235 -0.138	0.057 0.016 -0.041 -0.04 -0.846 -0.889 -0.014 -0.029	0.001 -0.077 -0.18 -0.066 -0.061 -0.163 -0.092 -0.093	0.033 0 -0.019 -0.007 -0.05 -0.032 -0.052 -0.023	-0.326 -0.227 -0.139 -0.102 0.098 -0.17 -0.151	-0.017 -0.096 -0.03 -0.082 -0.288 -0.156 -0.087	-0.00 -0.00 -0.00 -0.01 -0.04
ISAM JULES LPJ LPX OCN ORCHIDEE VISIT MMEM	-0.007 0.068 0.073 -0.027 0.004 0.02 -0.116 -0.006	-0.031 -0.045 -0.027 -0.009 -0.014 -0.014 -0.054 -0.029	-0.02 -0.021 -0.008 -0.007 -0.011 0.013 -0.048 -0.018	-0.044 -0.062 -0.028 -0.019 -0.026 -0.063 -0.073 -0.049	-0.065 -0.108 -0.012 0.073 -0.077 0.084 -0.042 -0.03	-0.021 -0.002 0.021 -0.005 0.006 0.007 -0.007 -0.004	-0.09 -0.887 -0.104 0.024 0.005 0.026 -0.119 -0.057	-0.02 -0.045 -0.229 -0.043 0.002 0.003 -0.199 -0.877	-0.028 0.102 -0.061 -0.009 -0.119 -0.11 -0.18 -0.066	0.012 0.015 -0.059 -0.035 -0.036 -0.101 -0.842 -0.834	-0.063 -0.081 -0.053 -0.067 -0.104 -0.077 -0.072	0.009 0.01 -0.028 -0.011 -0.018 -0.043 -0.043 -0.02	0.008 0.136 0.127 0.217 0.095 0.186 0.022 0.102	-0.045 -0.116 -0.041 0.02 -0.071 0.171 -0.117 -0.031	-0.003 0.092 0.002 -0.209 -0.239 -0.239 -0.254 -0.153	-0.037 -0.027 0.03 -0.015 -0.041 0.037 -0.128 -0.029	-0.022 0.206 0.076 0.046 -0.025 0.112 0.112	-0.014 -0.017 -0.038 -0.044 -0.06 -0.102 -0.062	-0.026 0.208 0.063 -0.051 -0.068 -0.323 -0.18 -0.072	-0.054 -0.163 -0.153 -0.095 -0.095 -0.172 -0.235 -0.138	0.057 0.016 -0.041 -0.04 -0.846 -0.846 -0.89 -0.014 -0.029	0.001 -0.077 -0.18 -0.066 -0.061 -0.163 -0.092 -0.093	0.033 0 -0.019 -0.007 -0.055 -0.032 -0.052 -0.023	-0.326 -0.227 -0.139 -0.102 0.098 -0.17 -0.151	-0.017 -0.096 -0.03 -0.082 -0.288 -0.156 -0.087	-0.00 -0.00 -0.00 -0.00 -0.04 -0.04

1 Figure 2 Trends in CO₂ seasonal peak-to-trough amplitude (AMP_{P-T}) (a) and trough-to-peak amplitude (AMP_{T-P}) (b) estimated by multi-model ensemble mean 2 (MMEM) under different scenarios at 26 northern temperate and boreal sites. We 3 present the results according to the latitudinal location of these sites. The single effect of 4 change in atmospheric CO2 ('CO2'), climate ('CLIM'), land use ('LU'), fossil fuel ('FF'), 5 ocean-air carbon flux ('Ocean') and wind ('Wind') on CO₂ seasonal amplitudes was derived 6 from transport simulation (T2 - T6), (T3 - T2), (T4 - T3), (T7 - T6), (T8 - T6) and T6, 7 respectively (see Methods and Table S4). For each scenario, we denote those significant (P <8 0.05) trends with two dots and marginal significant (P < 0.10) trends with a dot in the middle 9 of the bar. 10

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