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*Supplement of*

## **Contrasting effects of CO<sub>2</sub> fertilization, land-use change and warming on seasonal amplitude of Northern Hemisphere CO<sub>2</sub> exchange**

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TABLE S1 – Trends in  $SCA_{NBP}$  estimated by each inversion and the LSM multi-model ensemble mean (simulation S3) for different periods in the two selected latitudinal bands. CarboScope provides several inversion estimates with an increasing number of assimilated sites. For consistency, trends in  $SCA_{NBP}$  are estimated in each of the periods covered by the CarboScope inversions (s76 for 1980-2015, s85 for 1985-2015, and s93 for 1993-2015). To determine whether the changes to the LSMs and forcings in TRENDYv6 could have contributed to improving their estimates of the  $SCA_{NBP}$  trends, we compared the results to the TRENDYv4 multi-model ensemble mean (for a group of 9 models) for 1982-2013. Values are provided in  $Tg\ C\ yr^{-2}$  as in Figure 1, subscript and superscript values correspond to the 95% confidence interval limits of the trend fit and the asterisks indicate significant values ( $p < 0.05$ ). Because uncertainty in each dataset is calculated differently, we chose to present uncertainty in the trend only, but the footnotes indicate further details on uncertainty.

	$L_{>40N}$			$L_{25-40N}$				
Period	CAMS <sup>1</sup>	CarboScope <sup>2</sup>	TRENDYv6 <sup>3</sup>	CAMS <sup>1</sup>	CarboScope <sup>2</sup>	TRENDYv6 <sup>3</sup>		
1980-2015	17.3 <sup>21.8*</sup> <sub>12.9</sub>	13.3 <sup>16.6*</sup> <sub>10.1</sub> (s76)	9.5 <sup>12.8*</sup> <sub>6.1</sub>	-0.03 <sup>1.5</sup> <sub>-2.1</sub>	1.3 <sup>2.4*</sup> <sub>0.2</sub> (s76)	0.8 <sup>1.9</sup> <sub>-0.4</sub>		
1985-2015	16.2 <sup>21.9*</sup> <sub>10.6</sub>	11.7 <sup>16.7*</sup> <sub>6.7</sub> (s85)	10.6 <sup>15.1*</sup> <sub>6.2</sub>	-0.9 <sup>1.4</sup> <sub>-3.1</sub>	1.2 <sup>2.4*</sup> <sub>0.0</sub> (s85)	1.3 <sup>2.8</sup> <sub>-0.2</sub>		
1993-2015	19.5 <sup>28.3*</sup> <sub>10.7</sub>	19.2 <sup>27.5*</sup> <sub>10.9</sub> (s93)	14.4 <sup>21.9*</sup> <sub>6.9</sub>	-1.3 <sup>2.4</sup> <sub>-5.0</sub>	4.6 <sup>8.4*</sup> <sub>0.7</sub> (s93)	2.4 <sup>4.8*</sup> <sub>0.0</sub>		
	$L_{>40N}$				$L_{25-40Nca}$			
Period	CAMS	CarboScope	TRENDYv4 <sup>5</sup>	TRENDYv6 <sup>5</sup>	CAMS	CarboScope	TRENDYv4 <sup>5</sup>	TRENDYv6 <sup>5</sup>
1982-2013	14.6 <sup>19.8*</sup> <sub>9.3</sub>	11.9 <sup>15.7*</sup> <sub>8.2</sub> (s76)	6.7 <sup>9.0*</sup> <sub>4.5</sub>	9.6 <sup>13.8*</sup> <sub>5.4</sub>	-1.3 <sup>0.8</sup> <sub>-3.4</sub>	0.2 <sup>1.4</sup> <sub>-1.0</sub> (s76)	0.4 <sup>1.3</sup> <sub>-0.5</sub>	0.8 <sup>2.3</sup> <sub>-0.6</sub>

<sup>1</sup> The uncertainty only represents the uncertainty of the linear fit due to year-to-year variability. Values do not include uncertainty in CAMS SCA. Errors in NBP fluxes have been calculated for each latitudinal band ( $L_{>40N}$  and  $L_{25-40N}$ ) as  $1\sigma = \pm 0.1 PgC.yr^{-1}$ . Errors in  $SCA_{NBP}$  should be smaller since some terms should cancel out.

<sup>2</sup> The uncertainty only represents the uncertainty of the linear fit due to year-to-year variability. An estimate of the additional uncertainty from the inversion calculation can be obtained by comparing trends estimated for different sensitivity runs around CarboScope s85\_v4.1, shown in Figure S3.

<sup>3</sup> Values refer to the MEM only. SCA trends and respective error-bars for individual models are shown in Figure S4.

<sup>5</sup> Values refer to the set of models used here that also contributed to TRENDYv4 (CLM4.5, ISAM, JSBACH, JULES, LPX, ORCHIDEE/ORCHIDEE-MICT) and VISIT).

Table S2 – Statistical models tested to evaluate the best predictors of trends in  $SCA_{NBP}$ . N/A indicates that no significant fit has been found for a given number of predictors.

$N_{\text{Predictors}}$	Dataset	$L_{>40N}$			$L_{25-40N}$		
		Model	$R^2_{\text{adj}}$	AIC	Model	$R^2_{\text{adj}}$	AIC
1	CAMS	$SCA_{NBP} \sim CO_2 + c$	0.63	-35.6	N/A	N/A	N/A
	CarboScope	$SCA_{NBP} \sim CO_2 + c$	0.65	-42.7	$SCA_{NBP} \sim CO_2 + c$	0.09	-6.6
	MMEM	$SCA_{NBP} \sim CO_2 + c$	0.48	-19.5	N/A	N/A	N/A
2	<b>CAMS</b>	<b><math>SCA_{NBP} \sim T_{gs} + CO_2 + c</math></b>	<b>0.74</b>	<b>-47.8</b>	$SCA_{NBP} \sim Fert + CO_2 + c$	0.01	7.7
	CarboScope	$SCA_{NBP} \sim T_{gs} + CO_2 + c$	0.69	-46.0	$SCA_{NBP} \sim T_{gs} + CO_2 + c$	0.45	-23.8
	MMEM	$SCA_{NBP} \sim T_{gs} + CO_2 + c$	0.53	-22.0	$SCA_{NBP} \sim T_{gs} + CO_2 + c$	0.17	-10.6
3	CAMS	N/A	N/A	N/A	<b><math>SCA_{NBP} \sim Fert + T_{gs} + CO_2 + c</math></b>	<b>0.08</b>	<b>6.0</b>
	CarboScope	$SCA_{NBP} \sim A_{\text{Crop}} + A_{\text{For}} + CO_2 + c$	0.67	-42.6	<b><math>SCA_{NBP} \sim P_{gs} + T_{gs} + CO_2 + c</math></b>	<b>0.49</b>	<b>-25.9</b>
	<b>MMEM</b>	<b><math>SCA_{NBP} \sim A_{\text{For}} + T_{gs} + CO_2 + c</math></b>	<b>0.57</b>	<b>-24.7</b>	<b><math>SCA_{NBP} \sim WH + T_{gs} + CO_2 + c</math></b>	<b>0.33</b>	<b>-17.1</b>
4	CAMS	N/A	N/A	N/A	$SCA_{NBP} \sim Fert + Irrig + T_{gs} + CO_2 + c$	0.05	6.0
	<b>CarboScope</b>	<b><math>SCA_{NBP} \sim A_{\text{Crop}} + A_{\text{For}} + T_{gs} + CO_2 + c</math></b>	<b>0.71</b>	<b>-46.8</b>	N/A	N/A	N/A
	MMEM	N/A	N/A	N/A	N/A	N/A	N/A

Table S3 – References for the AMERIFLUX datasets used in Figure S8. Full references provided in the Supplementary Material references.

<b>Site</b>	<b>Ref</b>	<b>DOI</b>
<b>US-ARM</b>	Sebastien Biraud (2002-)	<a href="http://dx.doi.org/10.17190/AMF/1246027">http://dx.doi.org/10.17190/AMF/1246027</a>
<b>US-Bo1</b>	Tilden Meyers (1996-)	<a href="http://dx.doi.org/10.17190/AMF/1246036">http://dx.doi.org/10.17190/AMF/1246036</a>
<b>US-Br1</b>	John Prueger, Tim Parkin (2001-)a	<a href="http://dx.doi.org/10.17190/AMF/1246038">http://dx.doi.org/10.17190/AMF/1246038</a>
<b>US-Br3</b>	John Prueger, Tim Parkin (2001-)b	<a href="http://dx.doi.org/10.17190/AMF/1246039">http://dx.doi.org/10.17190/AMF/1246039</a>
<b>US-IB1</b>	Roser Matamala (2005-)	<a href="http://dx.doi.org/10.17190/AMF/1246065">http://dx.doi.org/10.17190/AMF/1246065</a>
<b>US-Ne1</b>	Andy Suyker (2001-)a	<a href="http://dx.doi.org/10.17190/AMF/1246084">http://dx.doi.org/10.17190/AMF/1246084</a>
<b>US-Ne2</b>	Andy Suyker (2001-)b	<a href="http://dx.doi.org/10.17190/AMF/1246085">http://dx.doi.org/10.17190/AMF/1246085</a>
<b>US-Ne3</b>	Andy Suyker (2001-)c	<a href="http://dx.doi.org/10.17190/AMF/1246086">http://dx.doi.org/10.17190/AMF/1246086</a>
<b>US-Ro2</b>	John Baker, Tim Griffis (2003-2017)	<a href="http://dx.doi.org/10.17190/AMF/1418683">http://dx.doi.org/10.17190/AMF/1418683</a>
<b>US-Twt</b>	Dennis Baldocchi (2009-2017)	<a href="http://dx.doi.org/10.17190/AMF/1246140">http://dx.doi.org/10.17190/AMF/1246140</a>

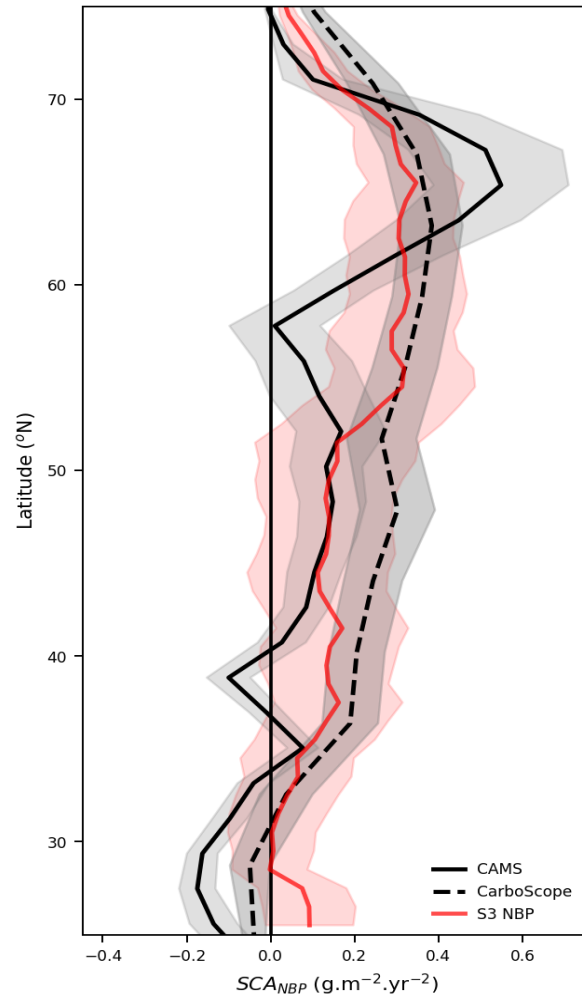


Figure S1 – Latitudinal distribution of  $SCA_{NBP}$  estimated by the two inversions and TRENDYv6 S3 (all forcings). The horizontal line separates the two latitudinal bands:  $L_{>40N}$  covering latitudes  $>40^{\circ}N$  (where both inversions estimate positive trends in  $SCA_{NBP}$ ), and  $L_{25-40N}$  covering the extra-tropical latitudes of the Northern Hemisphere ( $25-40^{\circ}N$ ). The shades indicate respectively the 95% confidence intervals for the trends in the case of inversions, and the model spread for LSMs.

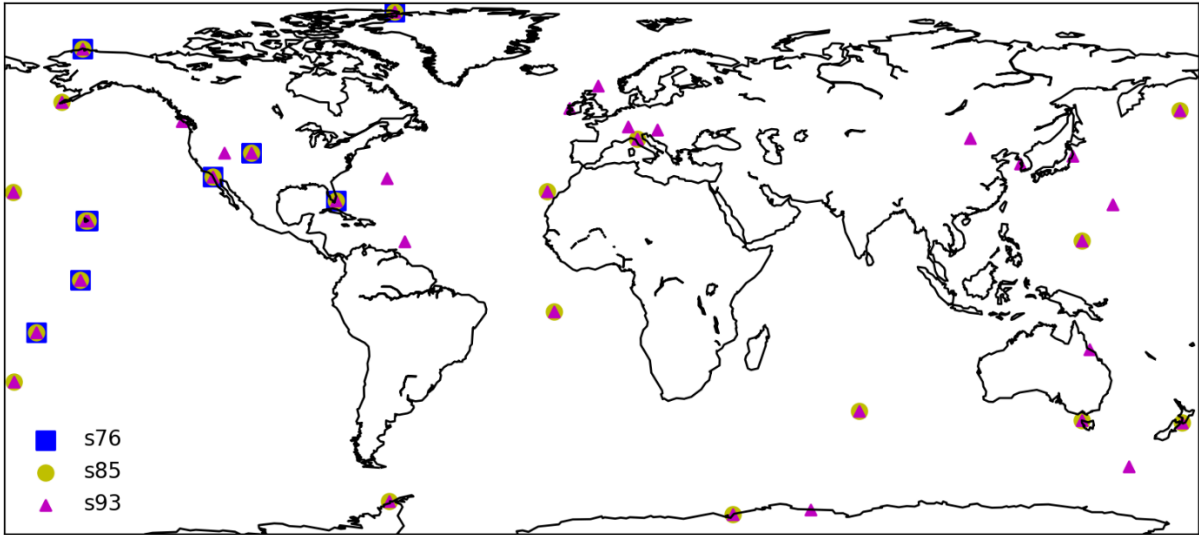


Figure S2 – Sites assimilated in the different versions of CarboScope v4.1 inversion: s76 (1976-2016), s85 (1985-2016) and s93 (1993-2016).

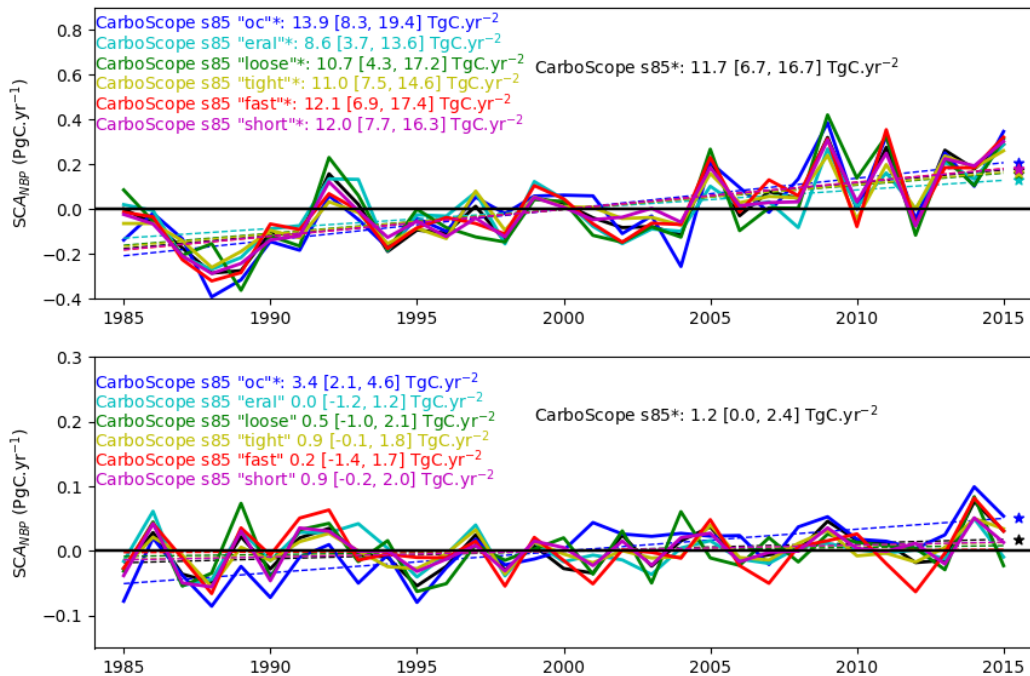


Figure S3 – Variability in SCA from different sensitivity tests around CarboScope s85, and respective trends. The different values provide an uncertainty range for s85\_v4.1 results; this range should not differ much for the other versions (i.e. s76, s93, Table S1). The different tests performed consisted of "oc": ocean fluxes fixed to pCO<sub>2</sub>-based estimates; "eral": forcing the transport model with fields from ERA-Interim reanalysis instead of NCEP; "loose" and "tight": scaling the a-priori sigma for all land and ocean flux components by 0.5 (dampening) and 2 (amplification), respectively; "fast": reducing the length of a-priori temporal correlations by a factor 2; "short": reducing the length of a-priori spatial correlations by a factor 3.

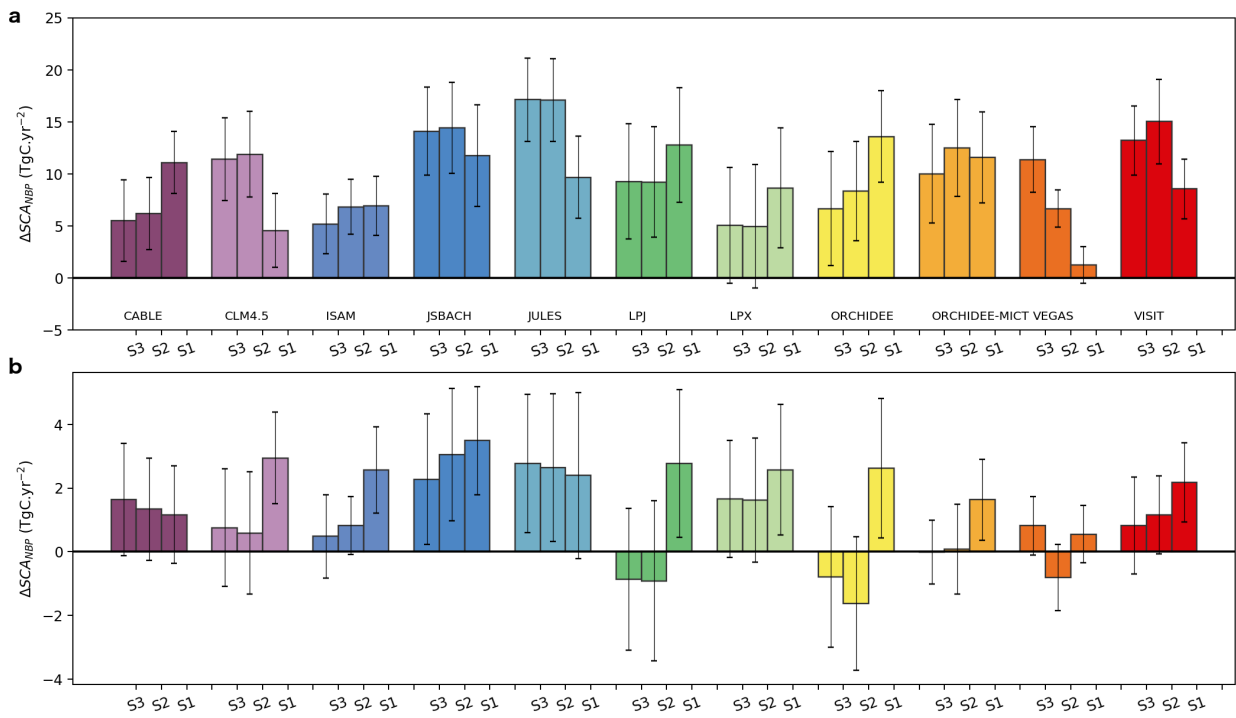


Figure S4 – Changes in SCA<sub>NBP</sub> estimated by individual models for each of the TRENDYv6 experiments for the two latitudinal bands a) L<sub>>40N</sub> b) and L<sub>25-40N</sub>). The error bars indicate the 95% confidence intervals for the trends. The labels in (a) indicate the corresponding model.

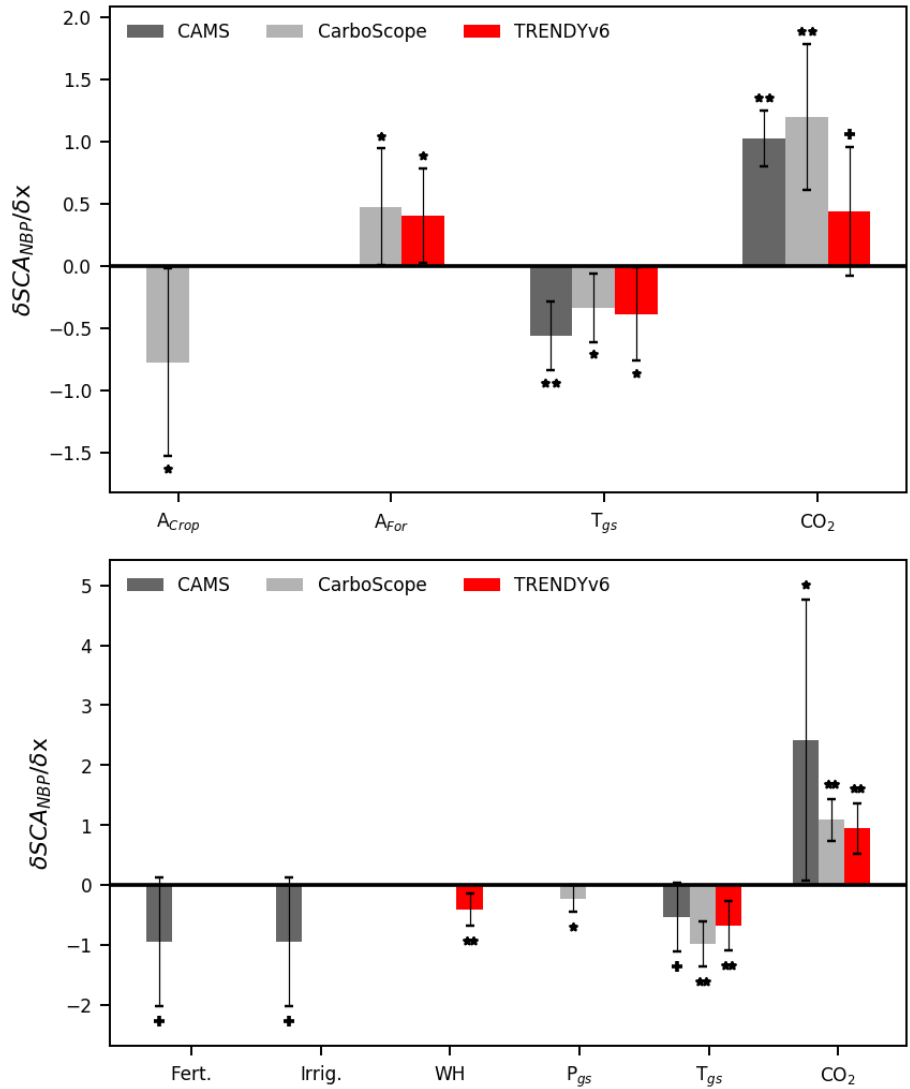


Figure S5 – Sensitivity of  $SCA_{NBP}$  to drivers for (a)  $L_{>40N}$  and (b)  $L_{25-40N}$ . The coefficients of the GLM fit of  $SCA_{NBP}$  shown here were used in Figure 3 to calculate the contribution of each variable to the trends in  $SCA_{NBP}$ .



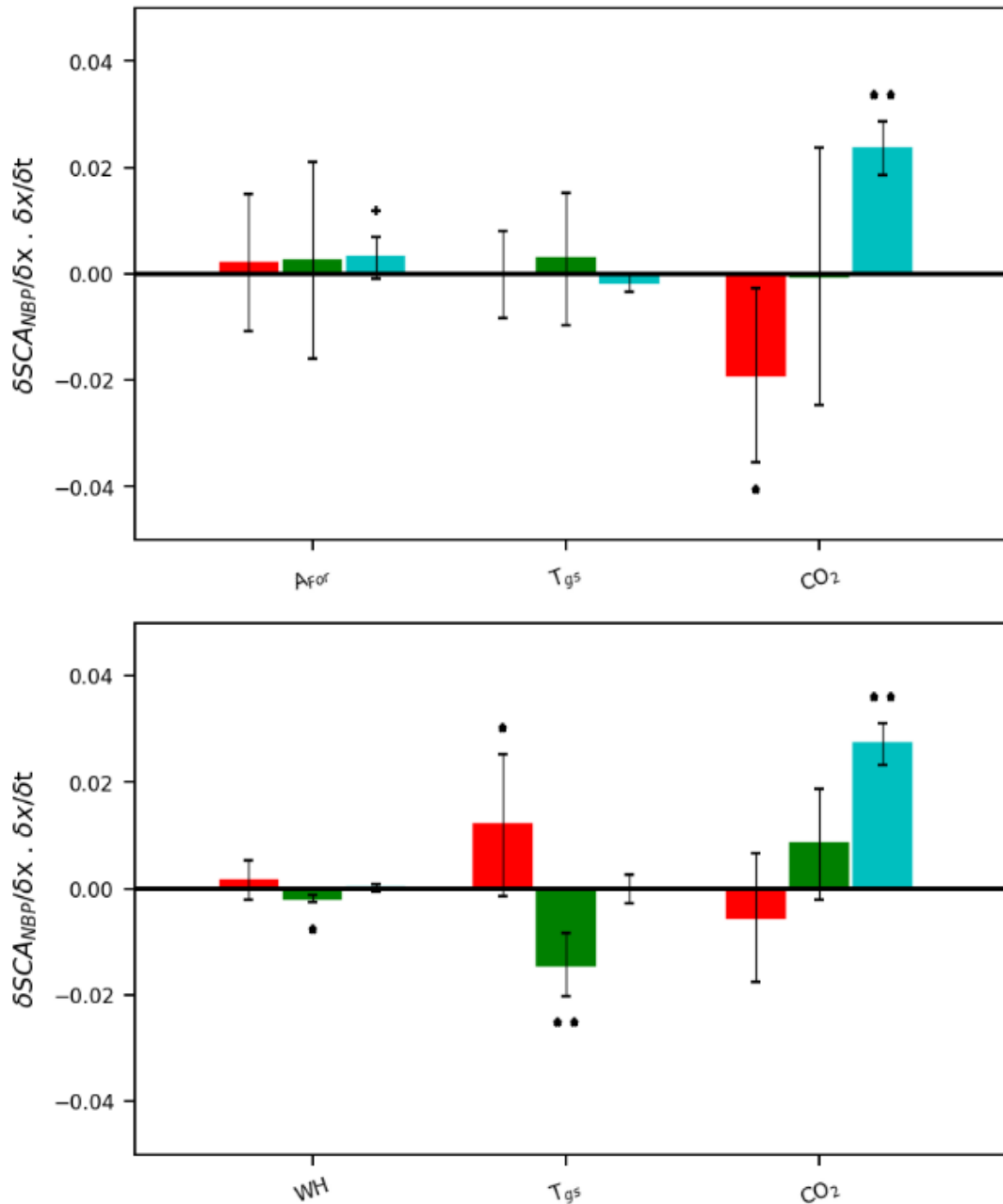


Figure S6 – Factorial verification of the drivers in TRENDY S3 for (a)  $>40^\circ N$  and (b)  $25-40^\circ N$ . The MLRM fit to the partial fluxes for the effects of LULCC (S3-S2, red), climate (S2-S1, green), and  $CO_2$  fertilisation (S1, cyan). Results should be compared to those in Figure 3. The significant predictors in the GLM fit to the LSMs in S3 should be detected in the corresponding factorial simulations. It should however be noted that management and fertilisation are already included in S1 and S2 for some models. The difference between S3 and S2 (LULCC effects) mainly suggest LULCC processes and does not identify the effect of  $CO_2$ , except if there are interactions between the  $CO_2$  fertilisation effect and LULCC emissions (e.g. higher emissions from deforestation because of higher C-stocks). The effect of  $CO_2$  is identified mainly by the difference in S1-S0 and S2-S1, possibly due to synergies between  $CO_2$  fertilisation and climate change. The effect of temperature should be evident in the difference between S2 and S1 (effects of climate), consistently found in L25-40N.

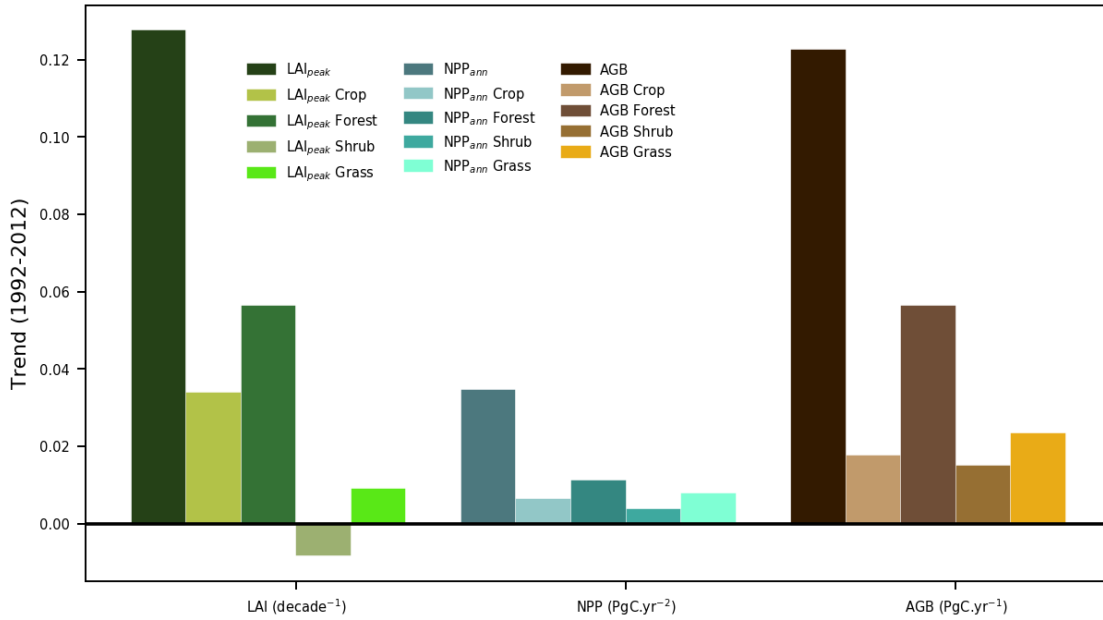


Figure S7 – Trends in vegetation growth for the land-cover types for 1992-2015. We analysed the trends in LAI (from GIMMS [Zhu et al., 2013]), NPP (based on AVHRR, from [Smith et al., 2016]), and AGB stocks (from passive-microwave satellite data for vegetation optical depth [Liu et al., 2016]). Trends in each variable were aggregated for the northern extra-tropical latitudes (>25°N, corresponding to L<sub>>40N</sub>+ L<sub>25-40N</sub>; left bar for each variable) and for four land-cover classes: cropland, forest, shrubland, and grassland. These classes were based on the ESA-CCI land-cover maps provided for 1992-2015, and each time interval was separately aggregated to account for changes in land-cover composition. The data sets covered distinct periods, so we selected periods of at least 20 years common to ESA-CCI LC and the vegetation data sets (1992-2012 for LAI, 1992-2011 for NPP, and 1993-2012 for AGB stocks).

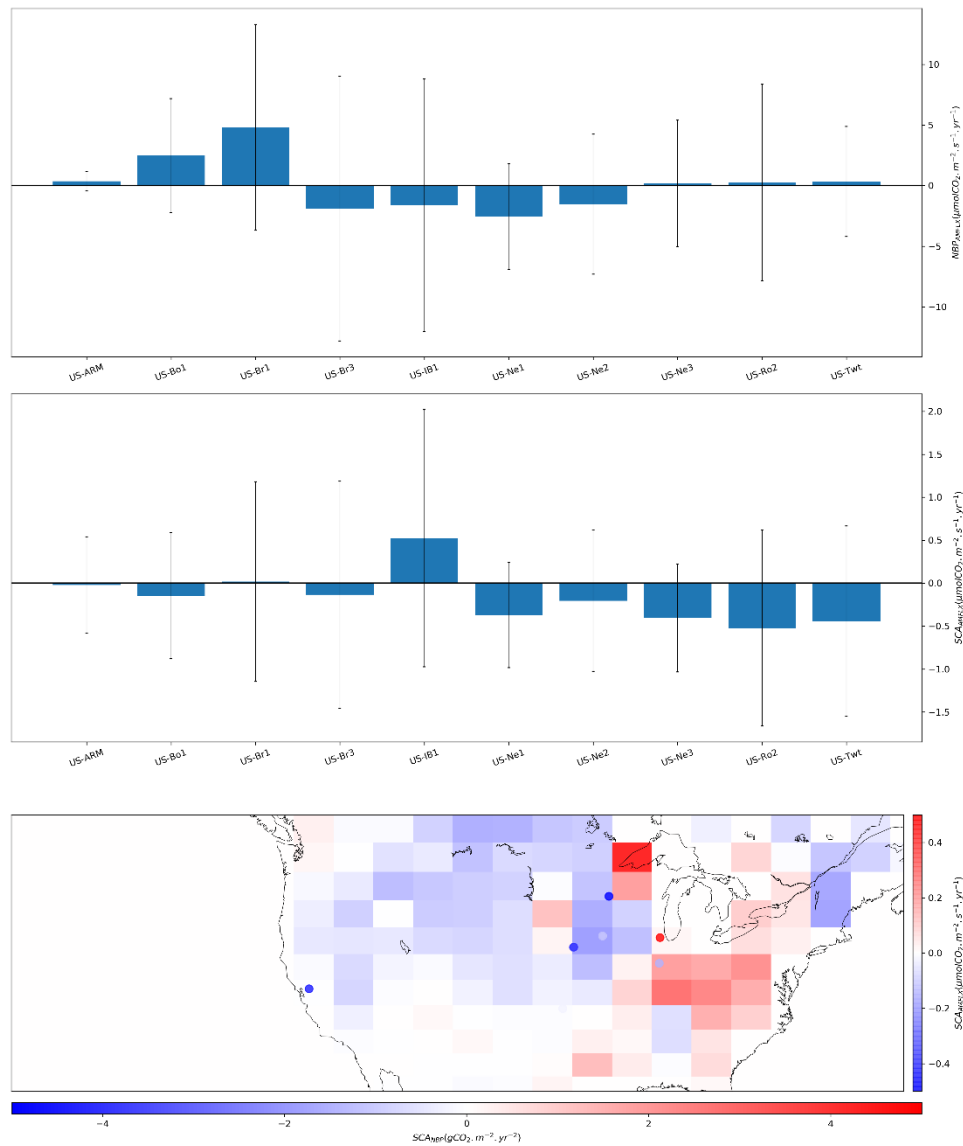


Figure S8 – Trends in the annual seasonal  $CO_2$  exchange in (a) AMERIFLUX cropland sites where eddy covariance was measured, and (b) comparison with the spatial patterns of CAMS. The trends in seasonal net ecosystem exchange were calculated for all AMERIFLUX sites in the continental USA (<https://ameriflux.lbl.gov/sites/site-search/#filter-type=all&igbp=CRO&region=USA>, Table S3). Data are provided at 30-min intervals during the various periods. Only sites with data for at least 5 years were selected, and the data were first filtered to remove noise and outliers by fitting annual Fourier functions (with four harmonics) to individual years to obtain a smooth seasonal cycle. The amplitude of the seasonal exchange was then defined as the difference between the maximum and minimum net land-to-atmosphere flux. The bars indicate the trends estimated for the corresponding validity periods (variable), and the error bars indicate the 95% confidence intervals. None of the sites has a significant trend (the error bars cross the x-axis), although most tend to be negative. The spatial pattern of the trends was then compared to the CAMS map, and CAMS also reports negative trends over most of midwestern USA where most areas are intensely cropped, consistent with the sign of the trends in the AMERIFLUX data.

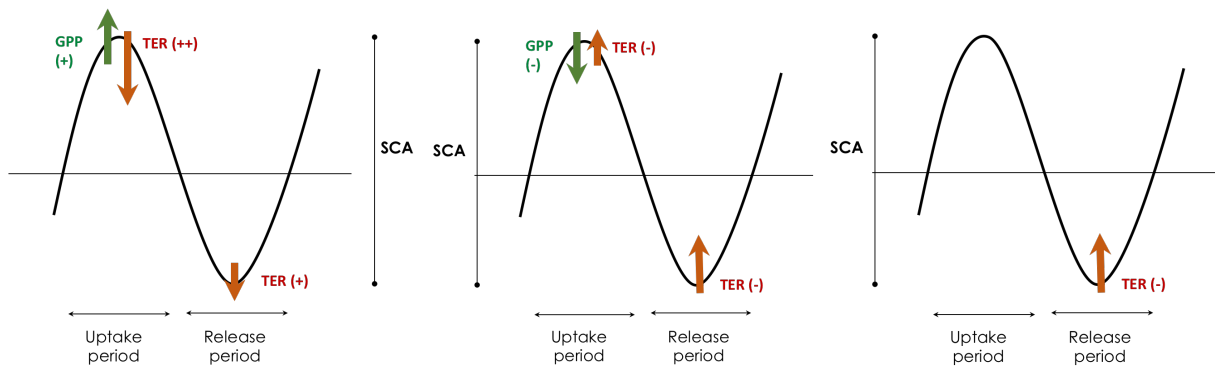


Figure S9: Conceptual scheme of the impacts of warming in  $SCA_{NBP}$ . The effects of temperature are seasonally-dependent as exemplified below. In the left panel, increasing T might increase GPP during the uptake period (having a positive effect on SCA), but at the same time increase maintenance respiration and decomposition in the uptake period, and leading to an increase in C available for decomposition during the release period. The latter effect would decrease SCA during the growing season, and increase SCA in the release period. On the other hand (middle panel), T might increase water-stress and contribute to decrease growing-season GPP (decreasing SCA). Consequently, maintenance respiration could be decreased in the growing-season (offsetting part of the GPP decrease effect), but also in the release-period (contributing to decrease SCA). Finally, in snow-covered regions, warming might contribute to reduce the snow-cover, which in turn might imply more TER during the growing-season but also increased GPP (left panel), but also have delayed effects during the release period, by reducing the insulation cover and inhibiting TER during the release period, which would contribute to a decrease in SCA (right panel).

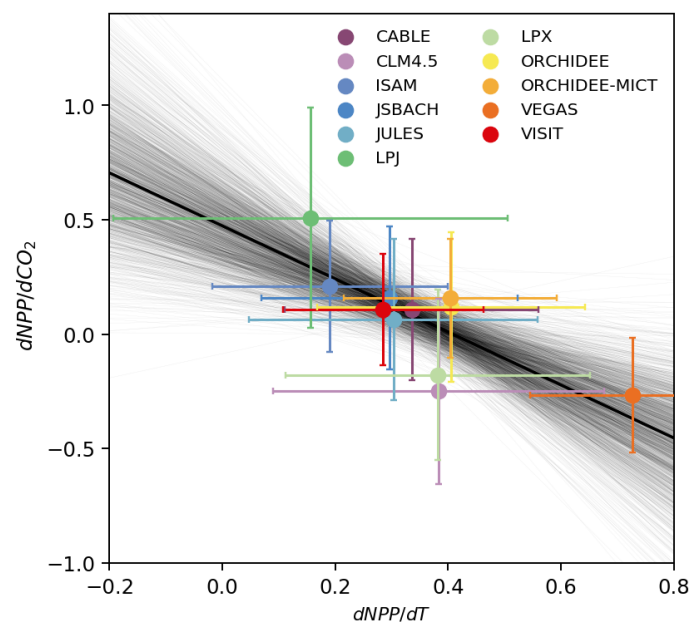


Figure S10 – Relationship between model sensitivity of net primary productivity (NPP) to  $CO_2$  ( $y$  axis) and to T ( $x$  axis). The distribution of the grey lines shows uncertainty in the relationship between the two variables.

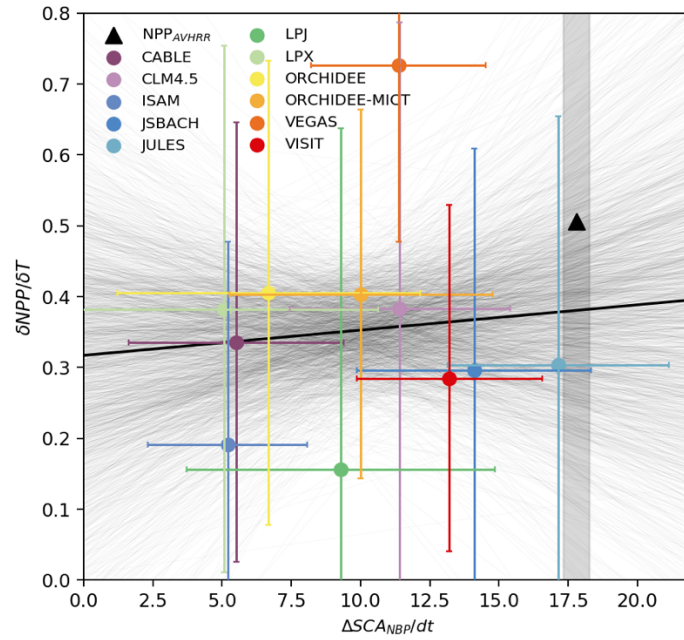


Figure S11 – Relationship between modelled trends in  $SCA_{NBP}$  and the sensitivity of (a) net primary productivity (NPP) to growing season temperature LSMs. The shaded areas indicate the inversion ranges for  $SCA_{NBP}$ . The distribution of the grey lines shows uncertainty in the relationship between the two variables.

## Supplementary References

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Andy Suyker (2001-)c AmeriFlux US-Ne3 Mead - rainfed maize-soybean rotation site, doi: 10.17190/AMF/1246086

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