

# Supplementary Information 5: Estimating losses due to CO<sub>2</sub> efflux

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*This pdf was generated from an Rmarkdown file, which includes all R code necessary to reproduce the estimations. The Rmarkdown file is available on github:  
<https://github.com/TTRademacher/Exp2017Analysis>.*

## Estimating total carbon losses due to CO<sub>2</sub> efflux

To estimate stem carbon losses due to CO<sub>2</sub> efflux, weekly CO<sub>2</sub> efflux measurements were made using a Li-Cor820 starting at 13.00h. The LiCor820 was attached to a PVC pipe (4" diameter) that were previously fitted and attached to the stem sections using silicone to create a closed chamber in which the enclosed air circulates via a small pump, similar to the Flux Puppy system (Carbone et al., 2019). CO<sub>2</sub> efflux and uncertainties were calculate from the change in CO<sub>2</sub> concentrations using the RespChamberProc package (Perez-Priego et al., 2015) in R (R Core Team, 2019). Stem sections were visited in the same order to reduce the role of diel variations over the course of the experiment. The order was initially randomised but slightly adjusted to assure that no treatment was systematically measured later in the day. Air temperatures and volumetric soil water content ranged between 7.1 and 23.9°C and 11.3 and 44.8% over the experiment, but less than 2.9°C or 0.1% on any particular measurement date. All data is publicly available on the Harvard Forest Data Archive (Rademacher and Richardson, 2020).

## Integration over the surface area of each stem section

Instantaneous fluxes of diffusive CO<sub>2</sub> loss through the bark were measured as described above and subsequently integrated across the surface area of each stem section to estimate the respiratrory loss in grams per stem section per day. To integrate the measured CO<sub>2</sub> efflux rates across the surface of each 10cm-stem section, we first calculated the surface area of the i-th stem section ( $A_{s,i}$ ) as follows:

$$A_{s,i} = \frac{cbh_{s,i}}{100} \times h$$

, where  $cbh_{s,i}$  is the circumference in centimetres, as measured in the field with a tape measure, and  $h$  is the height of the section in metres (here  $h = 0.1\text{m}$ ). The stem CO<sub>2</sub> efflux rate over this surface area ( $R_{i,k}$ ) was then determined by multiplication of the instantaneous CO<sub>2</sub> flux ( $f_{i,CO_2}$ ; in  $g\ m^{-2}\ day^{-1}$ ) and the surface area to get the CO<sub>2</sub> flux rate (in  $g\ day^{-1}$ ).

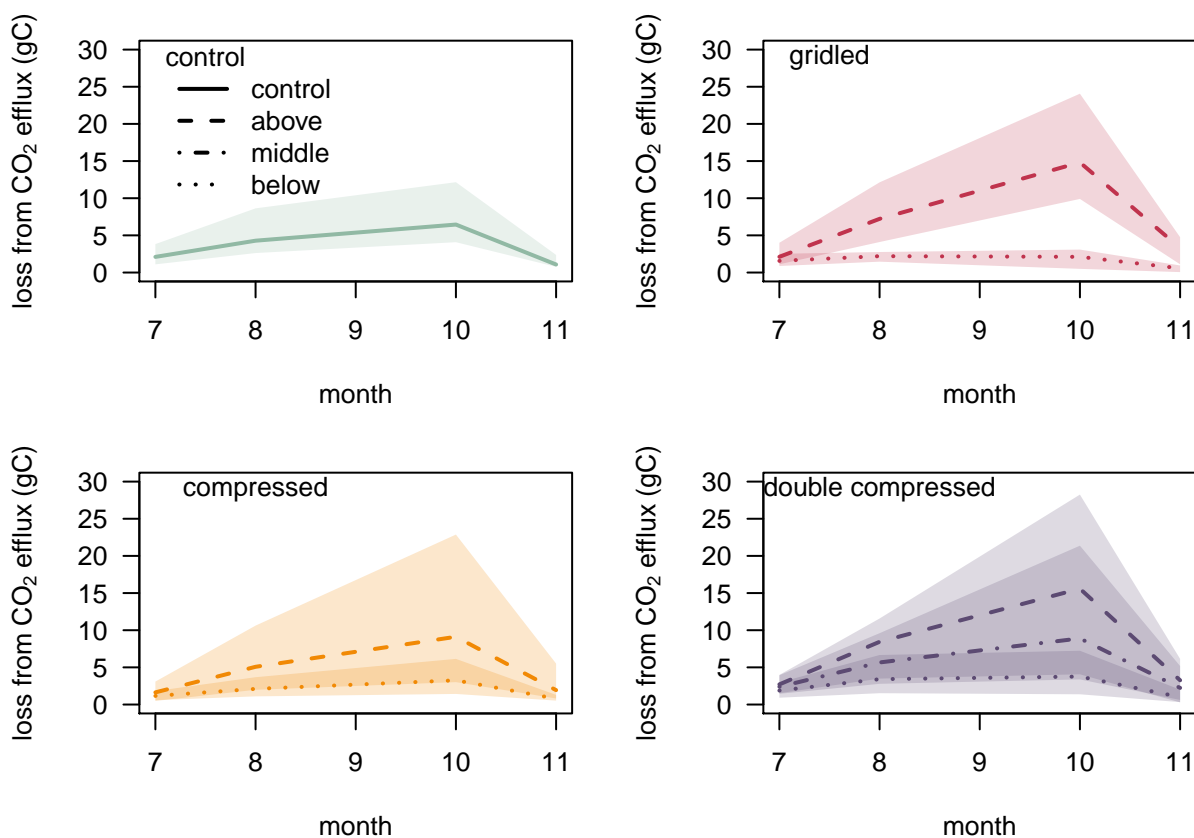
$$R_i = f_{i,CO_2} \times A_{s,i}$$

## Estimate respiratory losses for each periods

CO<sub>2</sub> flux rates were averaged for over four time periods sensu Figure 1 of the main text: (0) before experimental onset, (1) first month after start of the experiment, (2) second and third month after the start of the experiment, and (3) from the fourth month after the start of the experiment to late autumn. We then approximated the total loss of carbon for each combination of period  $j$  and stem section  $i$  ( $R_{i,j}$ ) by multiplying the temporal mean of spatially integrated CO<sub>2</sub> flux rates ( $R_i$ ; in  $g\ day^{-1}$ ) by the length of the period ( $l_p$  in days).

$$R_{i,j} = \frac{1}{n} \sum_{k=1}^n R_{i,k} \times l_p$$

, where  $n$  is the number of weekly sampling dates during each period. The figure below shows the average and standard deviation of the mean loss of carbon due to CO<sub>2</sub> efflux for each period (i.e., above the start date of each period).



## References

Carbone, M.S., Seyednasrollah, B., Rademacher, T.T., Basler, D., Le Moine, J.M., Beals, S., Beasley, J., Greene, A., Kelroy, J., Richardson, A.D., 2019. Flux Puppy – An open-source software application and portable system design for low-cost manual measurements of CO<sub>2</sub> and H<sub>2</sub>O fluxes. *Agricultural and Forest Meteorology* 274, 1–6.

Perez-Priego, O., Guan, J., Rossini, M., Fava, F., Wutzler, T., Moreno, G., Carvalhais, N., Carrara, A., Kolle, O., Julitta, T., Schruppf, M., Reichstein, M., Migliavacca, M., 2015. Sun-induced chlorophyll fluores-

cence and photochemical reflectance index improve remote-sensing gross primary production estimates under varying nutrient availability in a typical Mediterranean savanna ecosystem. *Biogeosciences* 12, 6351–6367.

Rademacher, T., Richardson, A.D., 2020. White Pine Girdling and Compression Experiment at Harvard Forest 2017-2019. Harvard Forest Data Archive.

R Core Team, 2019. R: A Language and Environment for Statistical Computing. R Foundation for Statistical Computing, Vienna, Austria.