# SENSITIVITY ANALYSIS OF A SOURCE PARTITIONING METHOD FOR H<sub>2</sub>O and CO<sub>2</sub> Fluxes via Large Eddy Simulations

Anne Klosterhalfena\*, Arnold F. Moeneb, Marius Schmidta, Todd M. Scanlonc, Harry Vereeckena and Alexander Grafa

- <sup>a</sup> Agrosphere Institute, IBG-3, Forschungszentrum Jülich GmbH, 52425 Jülich, Germany, www.fz-juelich.de; \*a.klosterhalfen@fz-juelich.de
- <sup>b</sup> Meteorology and Air Quality Group, Wageningen University and Research, 6708 PB Wageningen, the Netherlands
- <sup>c</sup> Department of Environment Sciences, University of Virginia, Charlottesville, VA 22904, United States

### 1 - INTRODUCTION

Eddy-covariance observations can only provide the *net* fluxes (of H<sub>2</sub>O and CO<sub>2</sub>) that emerge from plant canopies. However, for proper process understanding the contributions of transpiration, evaporation, photosynthesis, and respiration are required. Scanlon and Sahu (2008) and Scanlon and Kustas (2010) proposed a source partitioning method (**SK10**), that is based on: (1) high frequency raw data time series, (2) separate application of the flux-variance similarity theory to stomatal and non-stomatal components of the fluxes, and (3) assumptions on water use efficiency (WUE) on leaf-scale. We conducted Large Eddy Simulations (LES) for (a) contrasting canopy source/sink distributions, (b) varying relative magnitudes of soil sources and canopy sinks/sources (Fig.1), and (c) contrasting plant area density (PAD) distributions (affecting turbulence). SK10 was applied to the synthetic high frequency data and the effects of canopy type (PAD distribution), measurement height, sink-source-distributions, and varying assumed WUEs were tested regarding the partitioning performance. Here we focus on one PAD distribution (uniform).

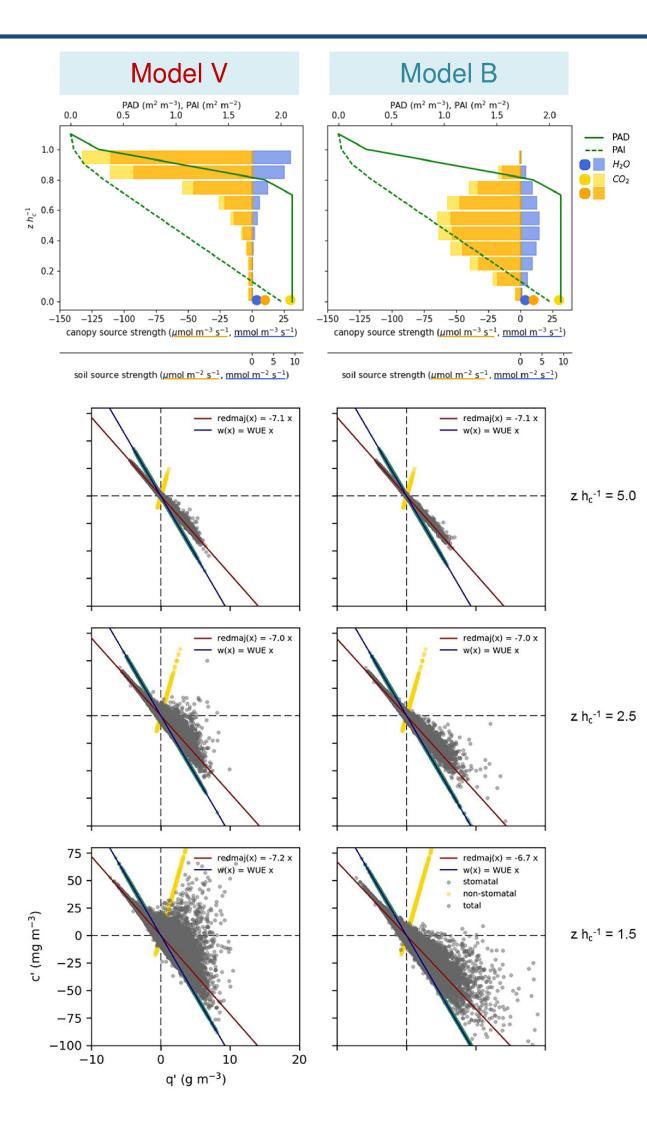
## 2 - LARGE EDDY SIMULATIONS

Model: DALES (Heus et al. 2010, Ouwersloot et al. 2016)

- Conditions: neutral
- 72 h<sub>c</sub> x 36 h<sub>c</sub> x 32 h<sub>c</sub> with 720 x 360 x 144 nodes
- grid resolution: 0.1  $h_c$
- simulation runtime: 720  $h_c u^{-1}$
- field sampling: every 6  $h_c u_*^{-1}$  for last 120  $h_c u_*^{-1}$
- PAI: 2 m<sup>2</sup> m<sup>-2</sup>, uniformly distributed
- 10 scalar sources in canopy and 1 soil source
- H<sub>2</sub>O and CO<sub>2</sub> fields generated from scalar fields by linear scaling based on source distribution

Fig. 1, top: Plant area density (PAD) distribution, cumulative plant area index (PAI) and variations of sink-source-distributions used to scale the LES scalar fields (*left*: ModelV after Sellers et al. 1992, *right*: ModelB after Ney et al. 2017), each with ten canopy sinks/sources (*bars*) and one soil source (*circle*). For CO<sub>2</sub>, two different soil sources were used, with the canopy sink adapted such that the net flux is the same.

Fig.2, bottom: Examples of sampled synthetic high frequency data of q' and c' at different 'measurement' heights for ModelV and ModelB (with the strong soil source). Differentiation between scalars originating from stomatal (green dots) and non-stomatal (yellow dots) processes. The blue line presents the WUE and the red line the reduced major axis regression between total q' and c'.



## 4 - RESULTS

- SK10 was able to approach the correct partitioning only for the strong soil source and observations within the roughness sublayer (maximum decorrelation) (Fig.4~top). The partitioning was sensitive to the parameterization of  $\rho_{c_p'c_{r'}}$  (transfer assumption, Fig.5). Replacing a parameterized  $\rho_{c_p'c_{r'}}$  with observed values (correcting the transfer assumption), the partitioning results improved and were realistic up to  $z \approx 3h_c$  (Fig.4~bottom).
- The partitioning results were very sensitive to the assumed WUE. This sensitivity was strongly modified by the quality of the estimation of the correlation  $\rho_{c_n'c_{r'}}$  (Fig.6).

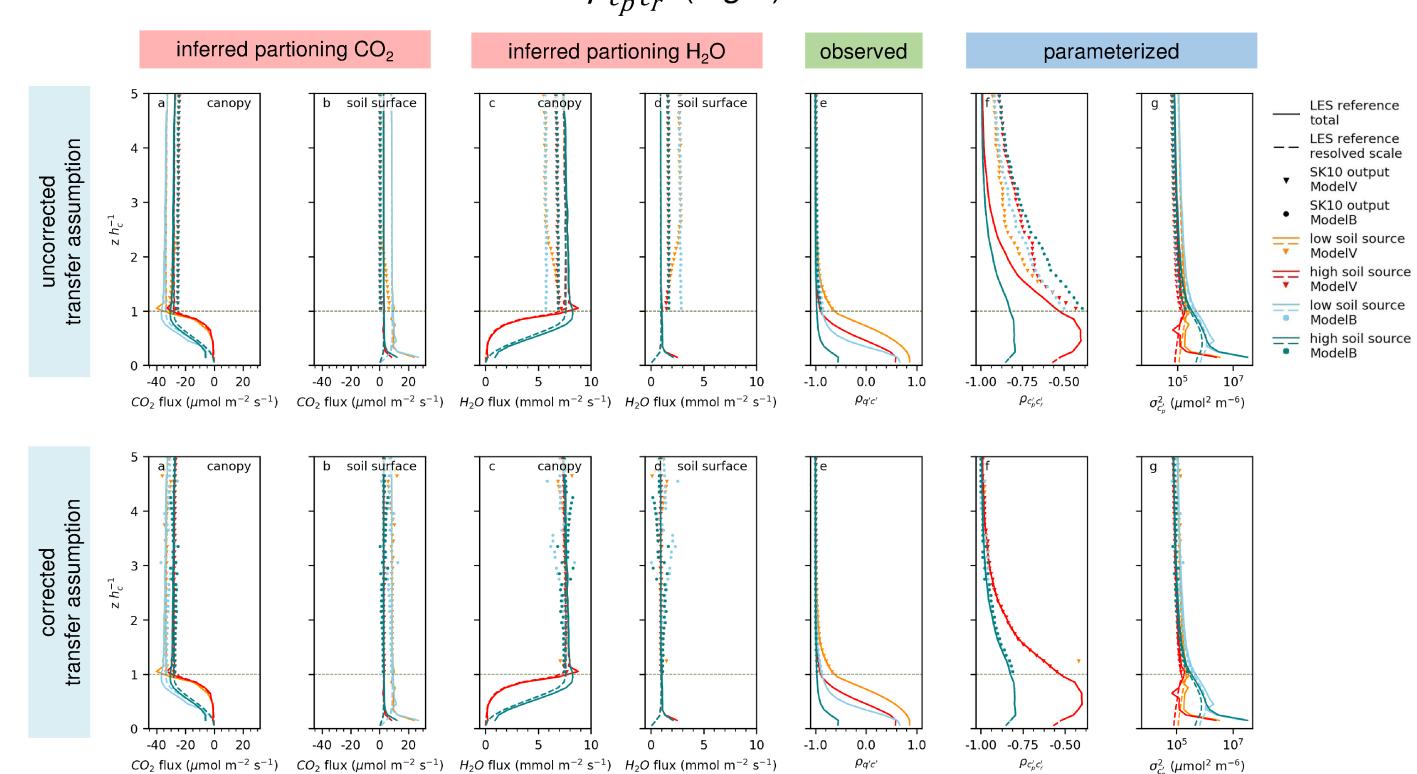


Fig.4: Vertical profiles of inferred  $H_2O$  and  $CO_2$  flux components, observed  $\rho_{q'c'}$  and parameterized  $\rho_{cp'c_{r'}}$  and  $\sigma_{cp'}$  for four sink-source-distributions (ModelV or ModelB, low or high soil source). Top: the partitioning results of SK10 with the original parametrization for  $\rho_{cn'c_{r'}}$ ; bottom: partitioning results with correction to the transfer assumption.

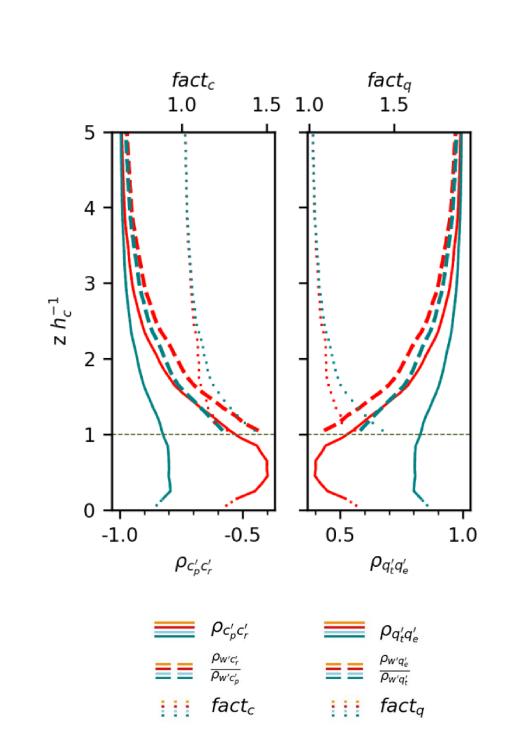


Fig.5: Comparison of real and parameterized correlation coefficients  $\rho_{c_p'c_{r'}}$  and  $\rho_{q_t'q_{e'}}$  (transfer assumption) and the corresponding correction factors ( $fact_a$ ,  $fact_c$ ).

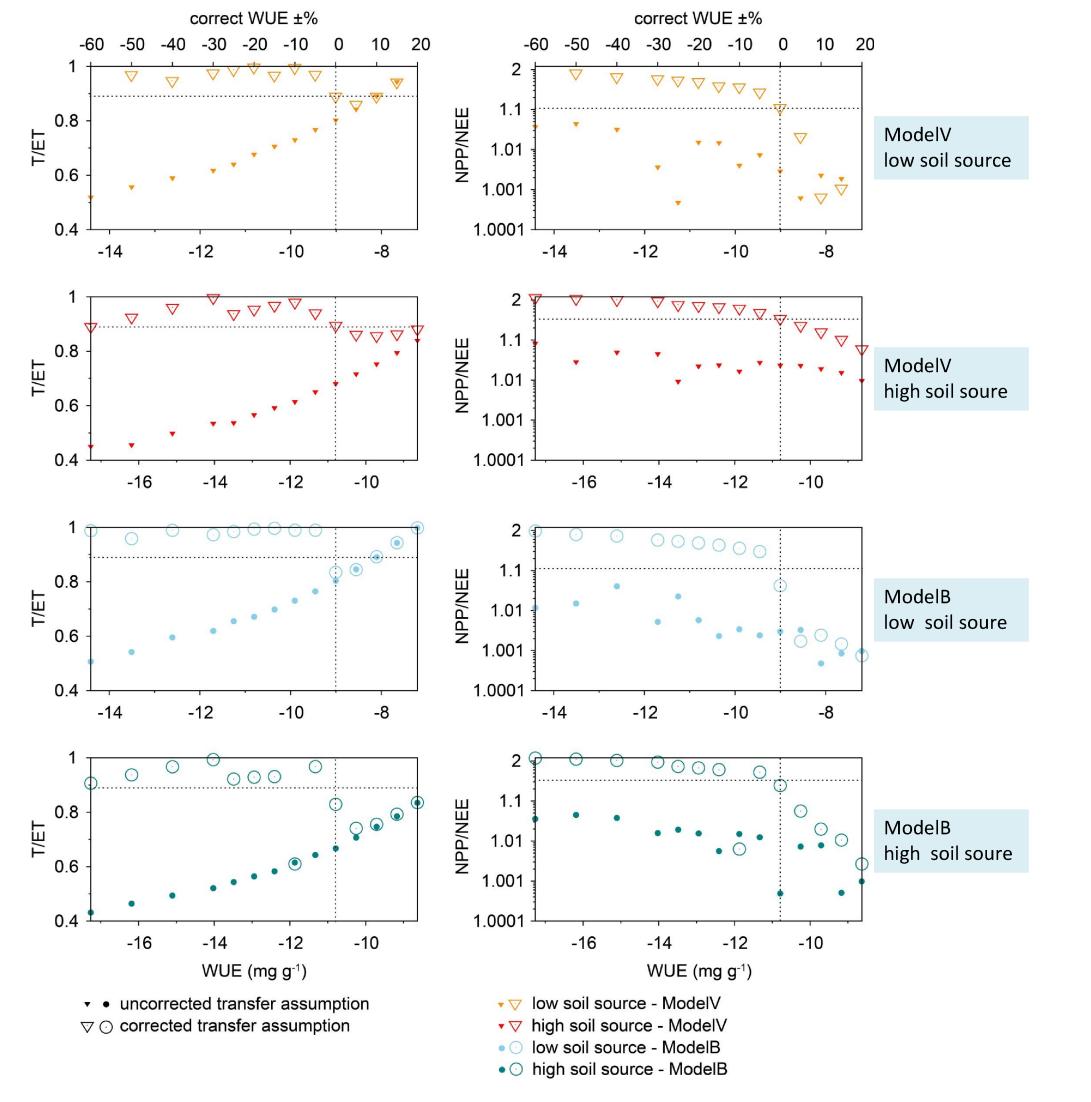


Fig.6: Results of partitioning fractions for  $H_2O$  (T/ET, left) and  $CO_2$  (NPP/NEE, right) fluxes in relation to the input WUE at a 'measurement' height of 2.5  $h_c$  with corrected and uncorrected transfer assumption. The true known imposed partitioning factors and WUE input are indicated by the dashed lines (T: transpiration, ET: evapotranspiration, NPP: net primary production, NEE: net ecosystem exchange).

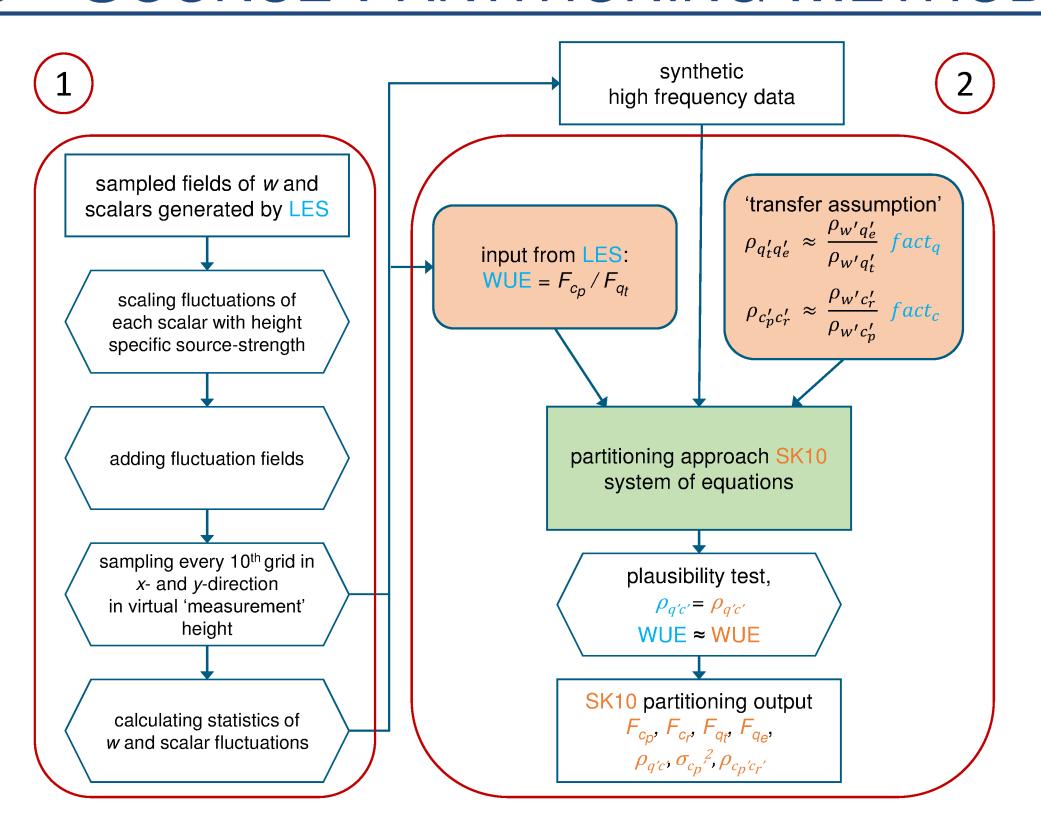
#### SPONSORED BY THE







## 3 - Source Partitioning Method



- Fig.3: Method: (1) generation of synthetic high frequency data with LES, (2) application of SK10 to derive the contributions to the fluxes of CO<sub>2</sub> and H<sub>2</sub>O.
- $(c_p: CO_2 \text{ related to photosynthesis}, c_r: CO_2 \text{ related to soil respiration}, F_x: flux of x, <math>h_c:$  canopy height,  $q_e: H_2O$  related to evaporation,  $q_t: H_2O$  related to transpiration,  $u_i:$  friction velocity at canopy top, z: height above soil surface)

## 5 - CONCLUSIONS

For a satisfying performance of SK10, a certain degree of decorrelation of q' and c' was needed: (1) enhanced by a clear separation between soil sources and canopy sinks/sources, (2) for observations within the roughness sublayer.

However, due to violation of the transfer assumption, the known true input WUE did not yield the known true input partitioning. This could only be achieved after introducing correction factors for the transfer assumption. However, it is unclear whether the profiles of these correction factors are universal and could be applied to field observations.

# ACKNOWLEDGMENT

SELLERS et al., 1992. Remote Sensing of Environment 42 (3), 187-216.

This research was supported by the German Federal Ministry of Education and Research BMBF, project IDAS-GHG (Grant 01LN1313A). The measurement infrastructure providing observational data was supported by the German Research Foundation DFG through the Transregional Collaborative Research Centre 32 (TR 32) and by the Helmholtz association through Terrestrial Environmental Observatories (TERENO). The computational facilities have been provided by the Netherlands Science Foundation under contract NWO 15774 (SH-312-15). The authors thank all technicians, engineers, and laboratory assistances in TR32 and TERENO for providing measurements of the test sites.

# REFERENCES

HEUS et al., 2010. Geoscientific Model Development 3 (2), 415-444.

KLOSTERHALFEN et al., 2019. Agricultural and Forest Meteorology 265, 152-170.

NEY et al., 2017. Poster presentation, EGU General Assembly 2017, Vienna, Austria, 23 Apr - 28 Apr 2017.

OUWERSLOOT et al., 2016. Boundary-Layer Meteorology 162 (1), 71-89.

SCANLON and KUSTAS, 2010. Agricultural and Forest Meteorology 150 (1), 89-99.

SCANLON and SAHU, 2008. Water Resources Research 44 (10), W10418, 15 pp.