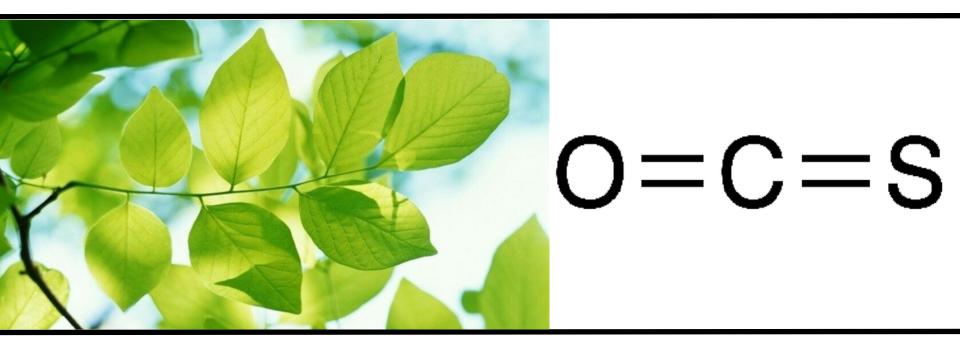
Inverse modelling with a coupled COS-CO<sub>2</sub> mixed-layer model Peter Bosman, Maarten Krol

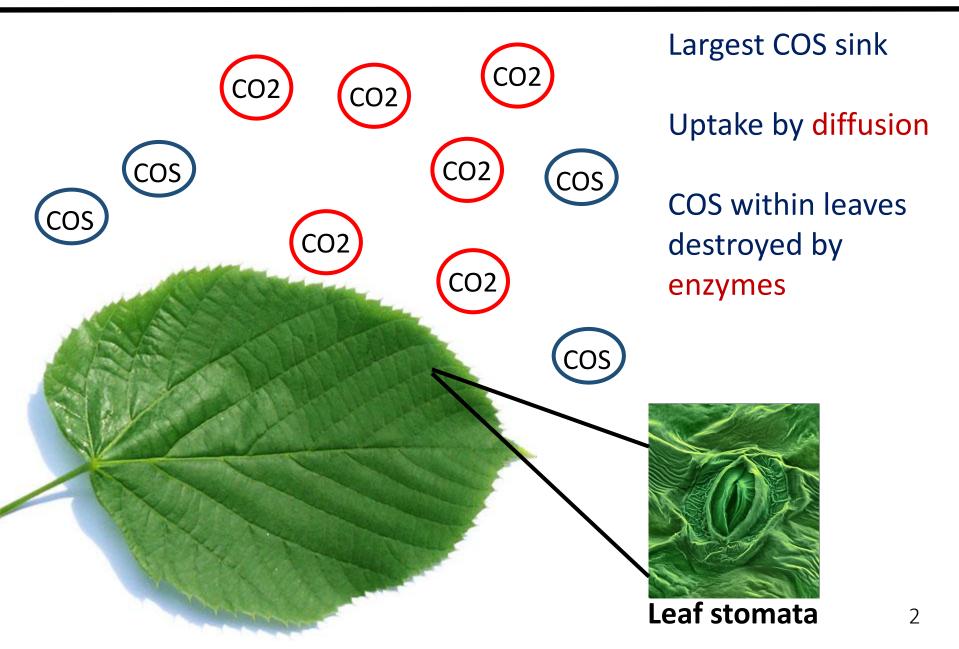


Overview of internship + current work in progress

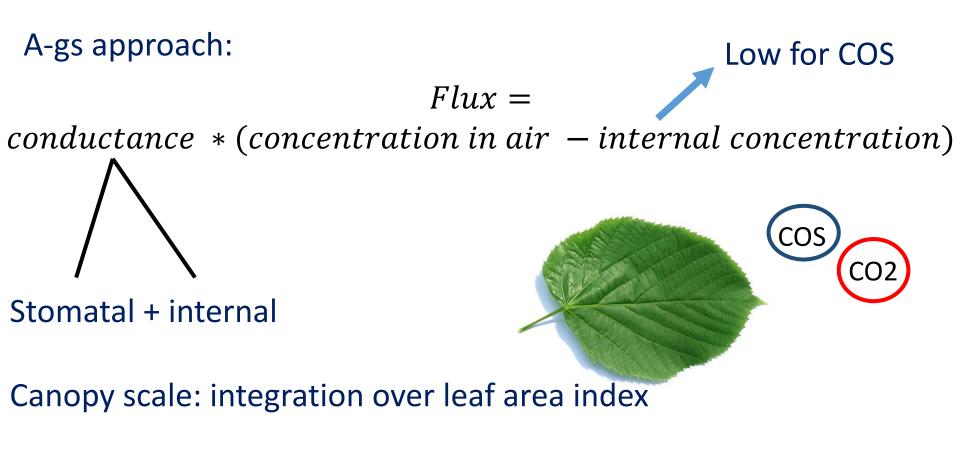


Jan 27, 2020

## Vegetation uptake



COS uptake and photosynthesis coupled to stomatal conductance -> crucial link between COS and photosynthesis

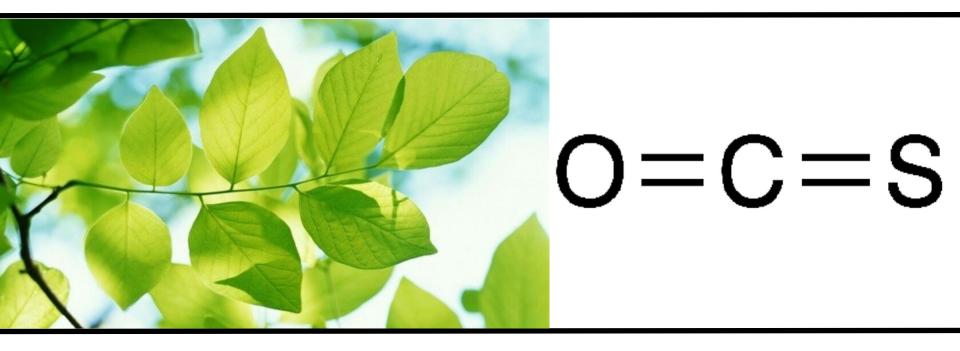


CO<sub>2</sub> and COS often coupled via ratio of deposition velocities, in this study coupled via conductance



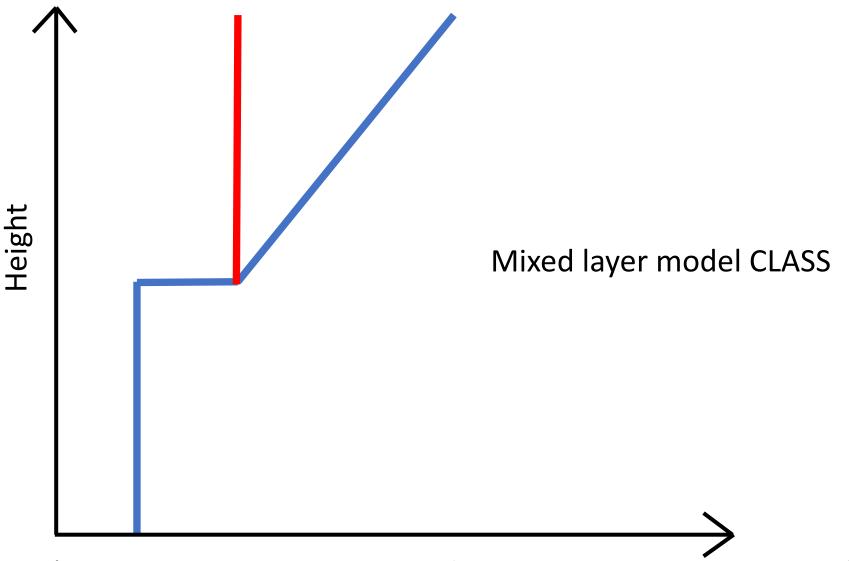


Overview of internship + current work in progress



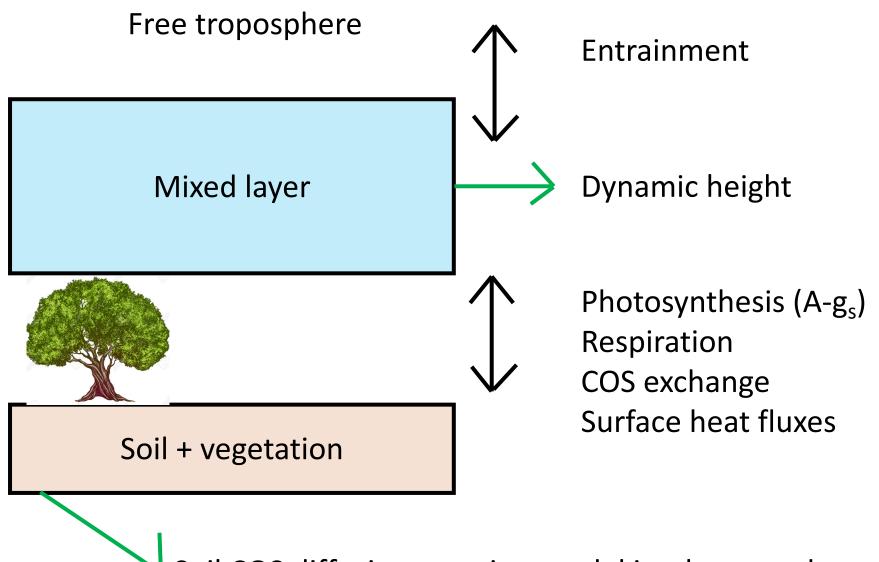
Jan 27, 2020

# The model



Scalar (potential temperature, specific humidity, chemical species) 5

# The model



Soil COS diffusion-reaction model implemented 6

## This research

Specific aim: Build an inverse modelling framework with a flexible cost function that allows for optimising different types of variables, including variables relating to the boundary layer dynamics

simple approach!



# The optimisation

(max) 4 parameters optimised in this study:

- alfa\_plant → scaling conductance influences COS and CO<sub>2</sub> uptake Main link with boundary layer dynamics removed
- alfa\_soil → scaling soil COS uptake/emission influences COS uptake only
- FTC\_COS → free tropospheric concentration of COS
- FTC\_CO2\_scale → scale for free tropospheric concentration of CO<sub>2</sub>

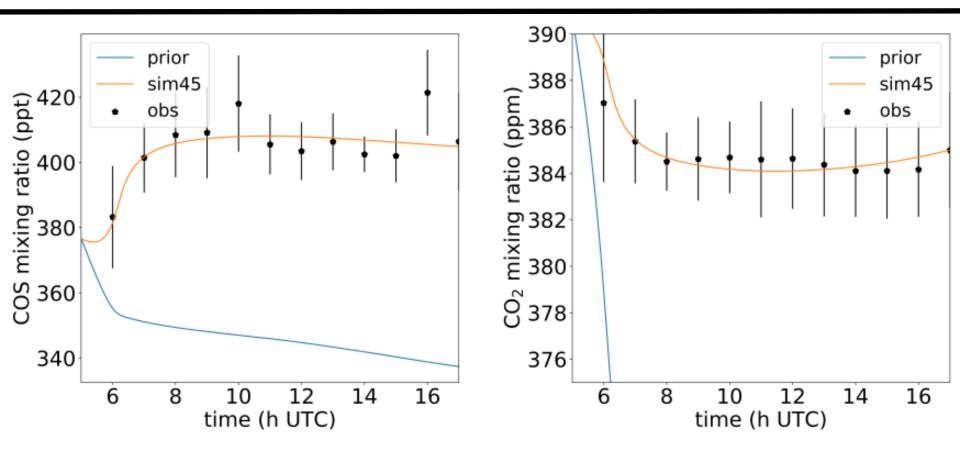
Dataset from boreal forest in Finland – Linda Kooijmans

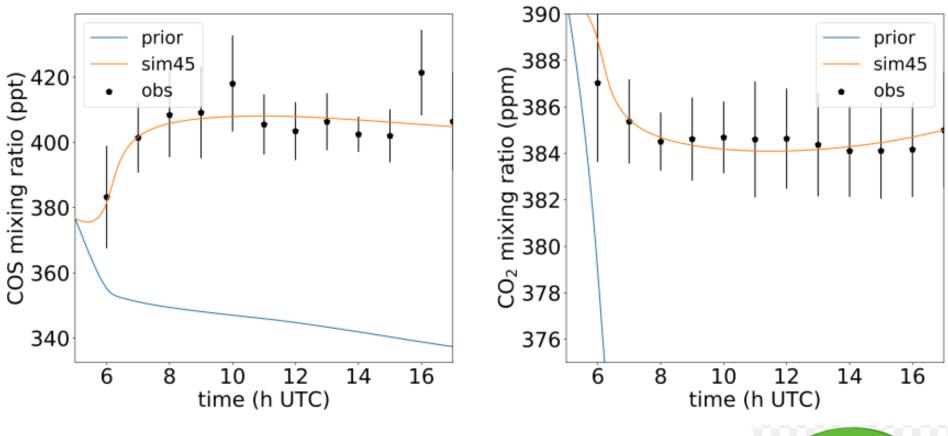
COS and CO<sub>2</sub> mixing ratios at 125 m eddy-covariance fluxes at 23 m

+...

Averaged over 7 sunny August days

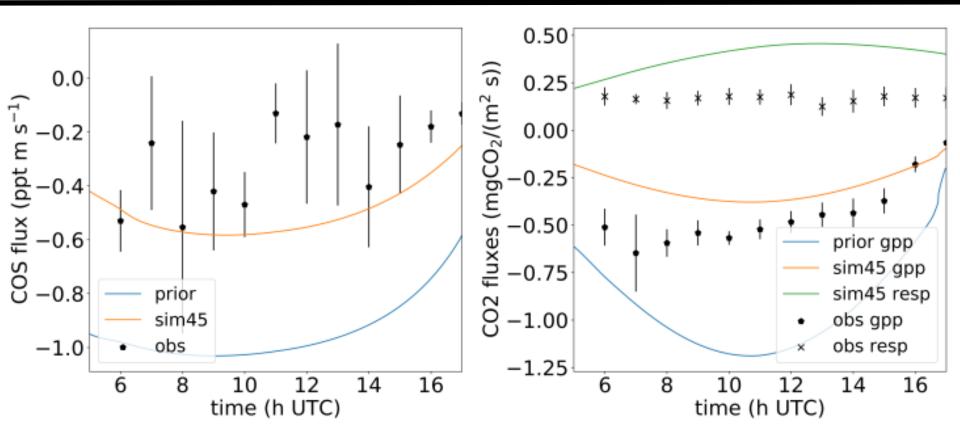






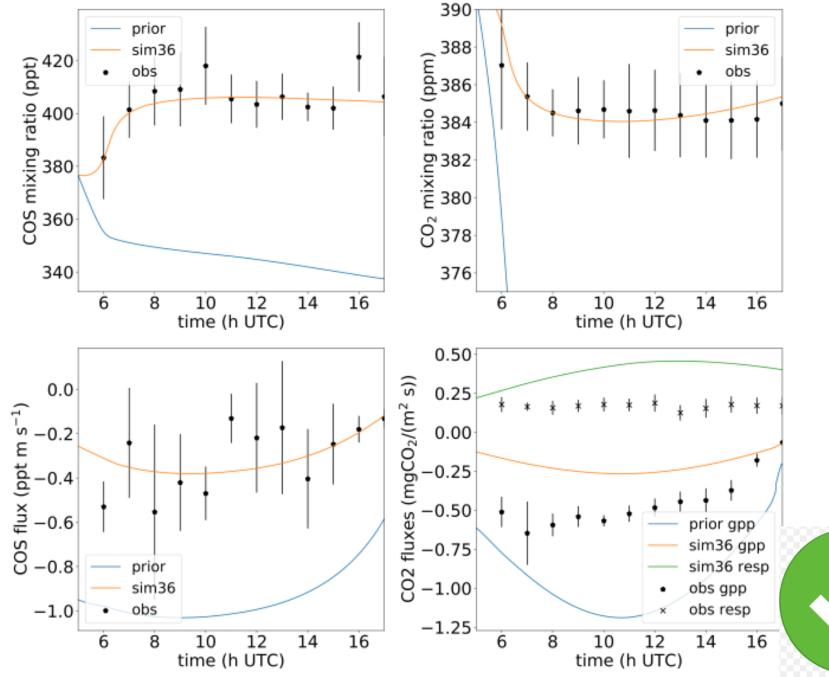
Cost function prior:2067.45Cost function optimised:3.95



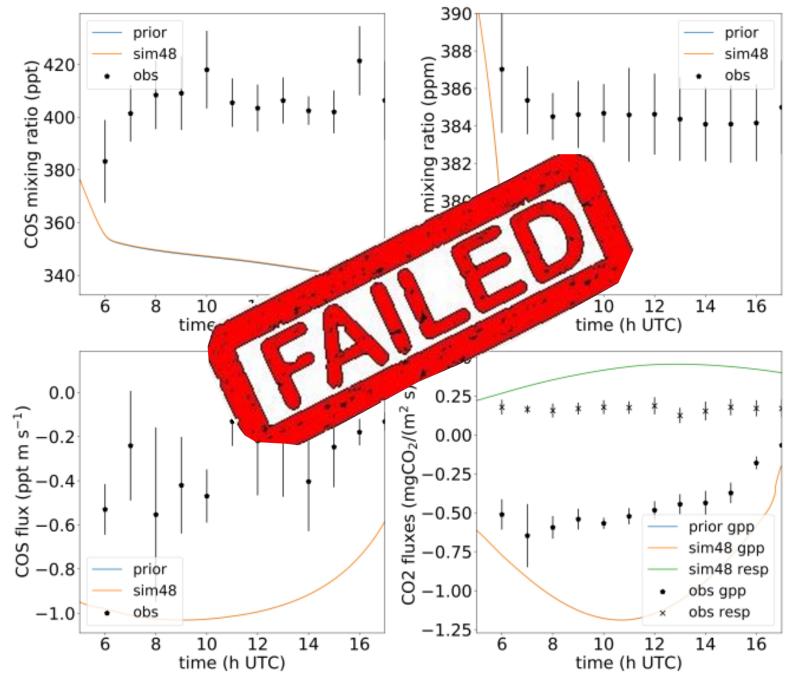


Fluxes can be improved --> Add to cost function!

#### With net COS flux in cost function:



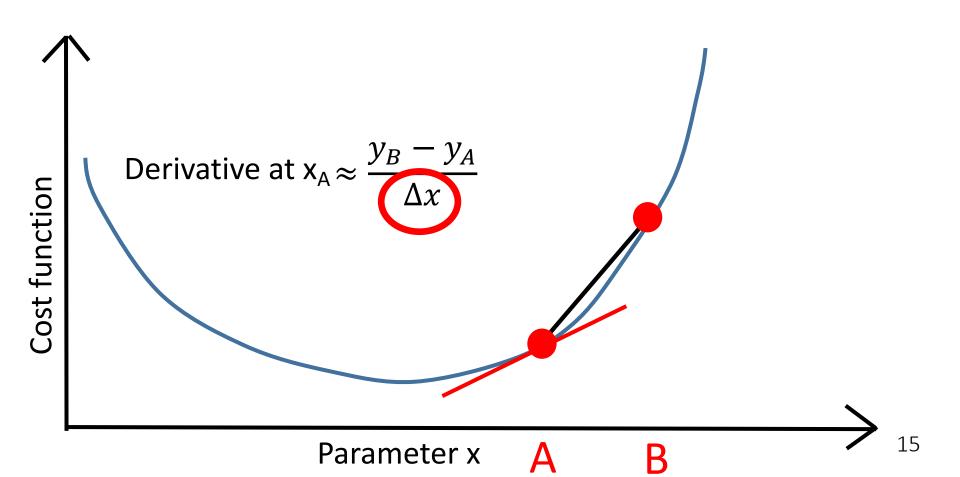
#### With net COS flux and gpp in cost function:



14

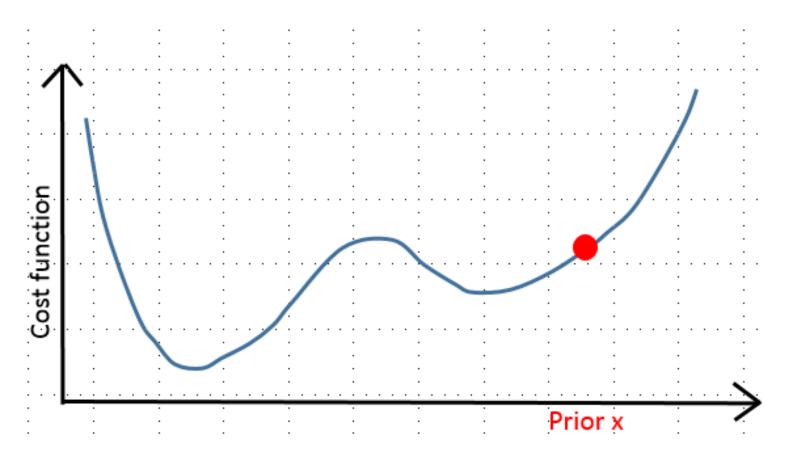
# Challenges

Derivative is approximated numerically (forward perturbation only) Analytical derivative requires construction of the *adjoint model* 



# Challenges

Derivative is approximated numerically (forward perturbation only) Analytical derivative requires construction of the *adjoint model* Model is non-linear



# Benefits of the framework

Cost function can contain any variable with observations

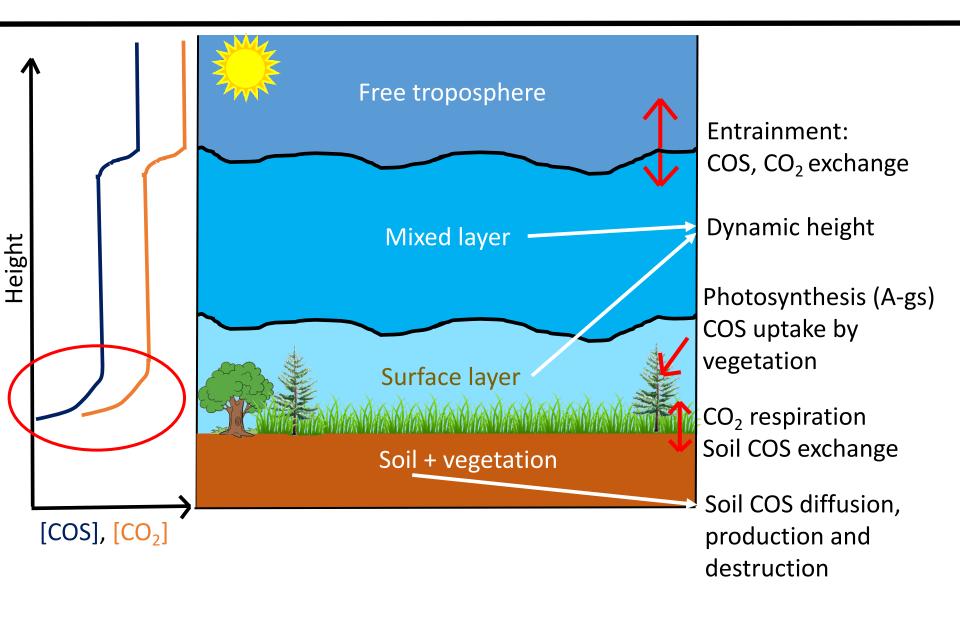
Any parameter can be optimised

Future goal: No more messing with manual parameter fitting!!

Challenges remain
→ Switch to analytical derivative?

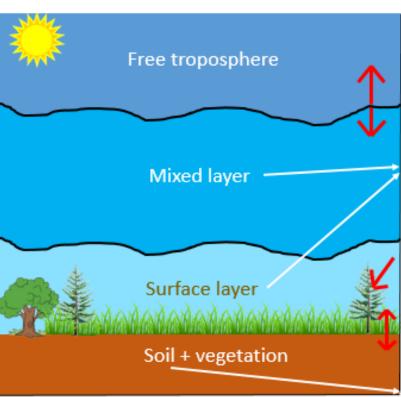


### Current work



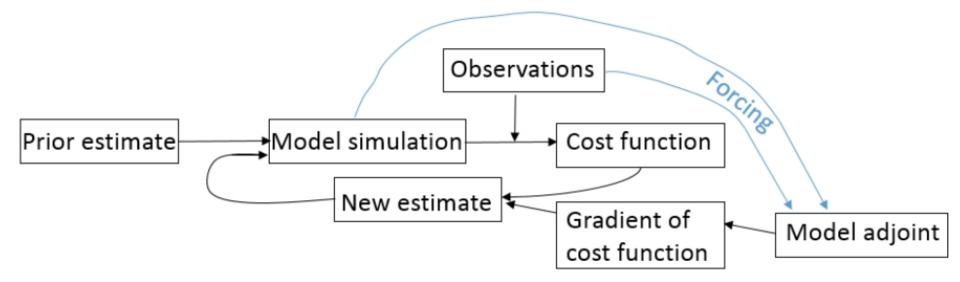
# Novelties and challenges

- In between ecosystem and global study
- Incorporate several type of obs
- Strongly nonlinear model
- Parameter dependency Cost function Prior



#### Current work

Analytical derivative : construction of the mixed-layer adjoint



## Adjoint modelling

Model code: C = 3 \* A + 5 \* B

Tangent linear model code:

dC = 3 \* dA + 5 \* dB $\begin{bmatrix} dC \\ dB \\ dA \end{bmatrix} = \begin{bmatrix} 0 & 5 & 3 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} dC \\ dB \\ dA \end{bmatrix}$ Transpose matrix:

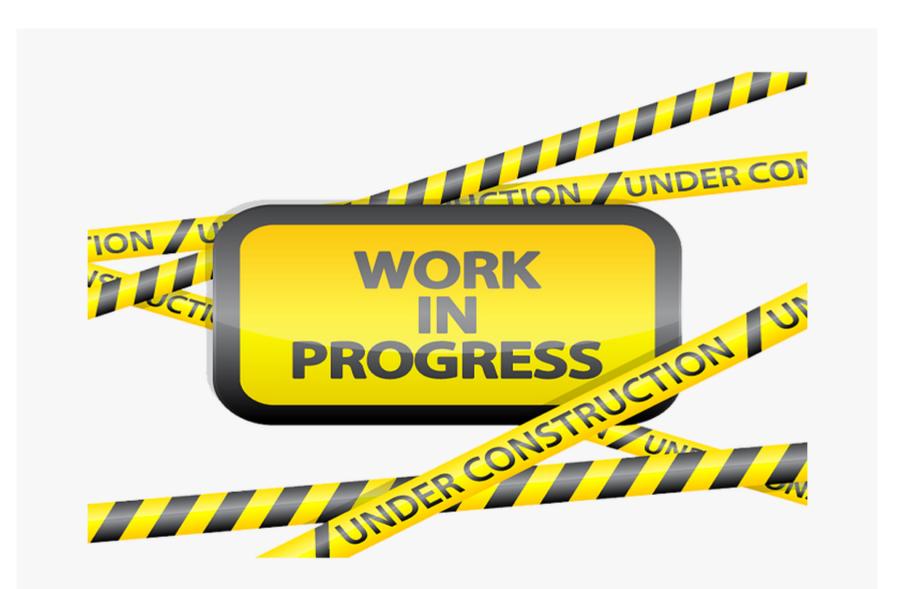
[adC]		0	0	[0	[adC]
adB	=	5	1	0	adB
ladA		3	0	1	LadA]

Adjoint model code: adA = 3 \* adC + adA adB = 5 \* adC + adBadC = 0

# Adjoint model code example

```
#statement ddeltagtend = model.gammag * (dwe + dwf_ddeltatheta - self.dM)
# - dqtend + dw q ft dh
self.adwe += model.gammag * self.addeltagtend
self.adwf_ddeltatheta += model.gammaq * self.addeltaqtend
self.adM += -model.gammag * self.addeltagtend
self.adgtend += - self.addeltagtend
self.adw g ft dh += self.addeltagtend
self.addeltagtend = 0
#statement ddeltathetatend = model.gammatheta * (dwe + dwf_ddeltatheta - self.dM)
#- dthetatend + dw th ft dh
self.adwe += model.gammatheta * self.addeltathetatend
self.adwf ddeltatheta += model.gammatheta * self.addeltathetatend
self.adM += -model.gammatheta * self.addeltathetatend
self.adthetatend += - self.addeltathetatend
self.adw theta ft dh += self.addeltathetatend
self.addeltathetatend = 0
#statement dCOStend = (self.dwCOS - dwCOSe - self.dwCOSM) / h + (wCOS - wCOSM)
#* (-1) * h**(-2) * self.dh + self.dadvCOS
self.adwCOS += 1 / h * self.adCOStend
self.adwCOSe += - 1 / h * self.adCOStend
self.adwCOSM += - 1 / h * self.adCOStend
self.adh += (wCOS - wCOSe - wCOSM) * (-1) * h**(-2) * self.adCOStend
self.adadvCOS += self.adCOStend
self.adCOStend = 0
```

## Adjoint modelling

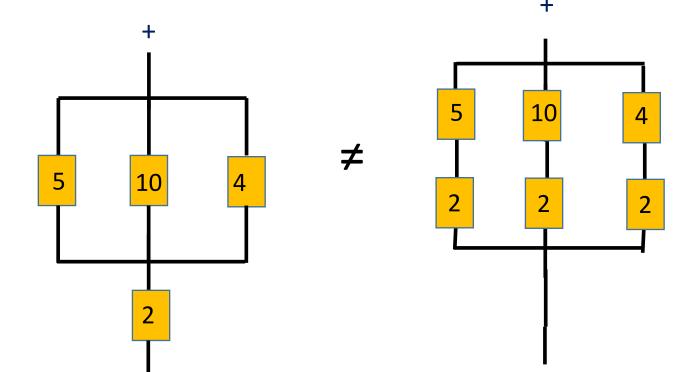


## Extra slides

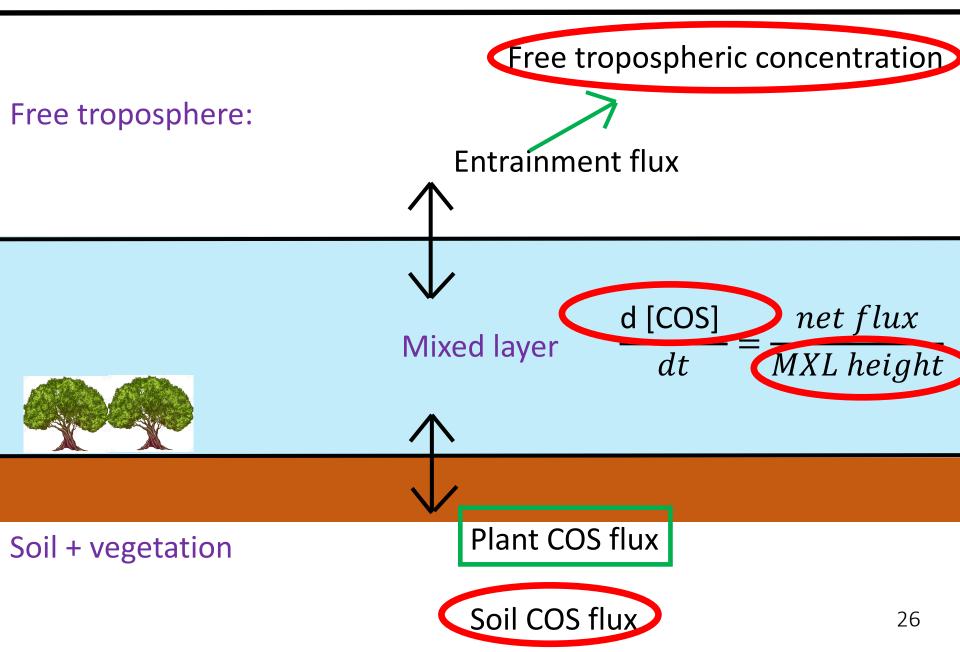
### COS conductance issues

Assume canopy consists of three leaves of same size, different stom resistance And C\_air = 10

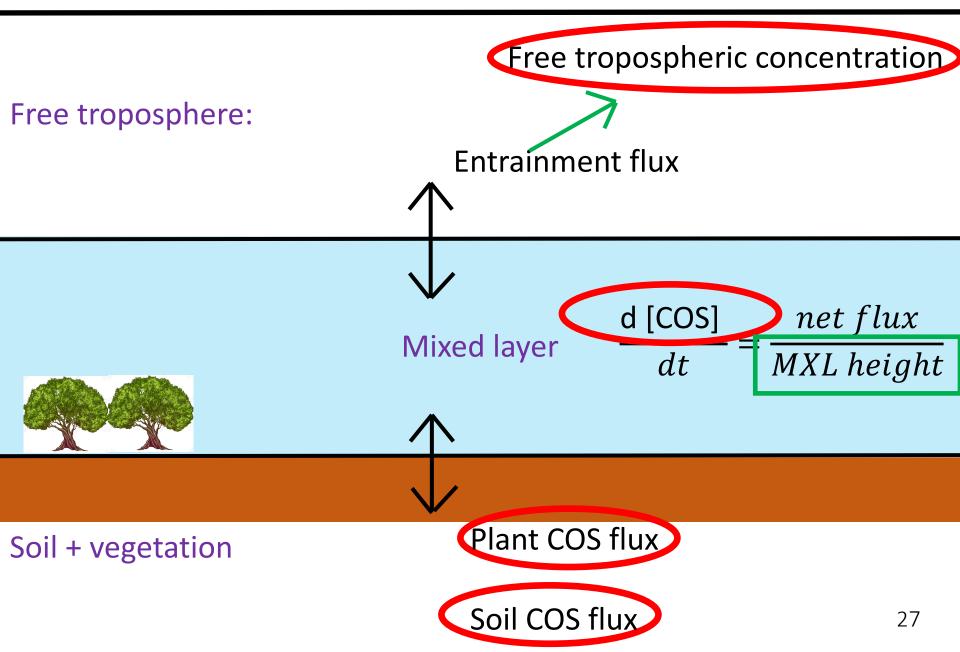




## Potential of the framework

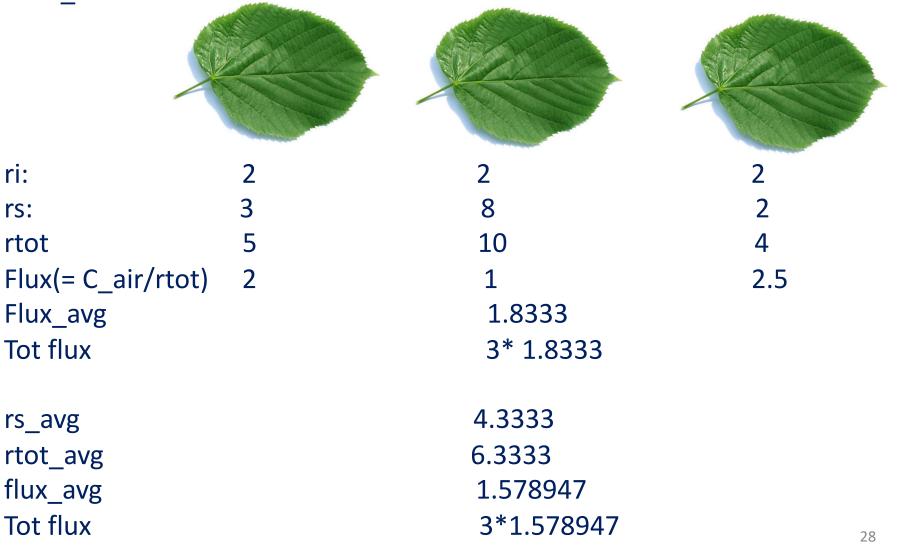


## Potential of the framework



## COS conductance issues

Assume canopy consists of three leaves of same size, different stom resistance And C\_air = 10



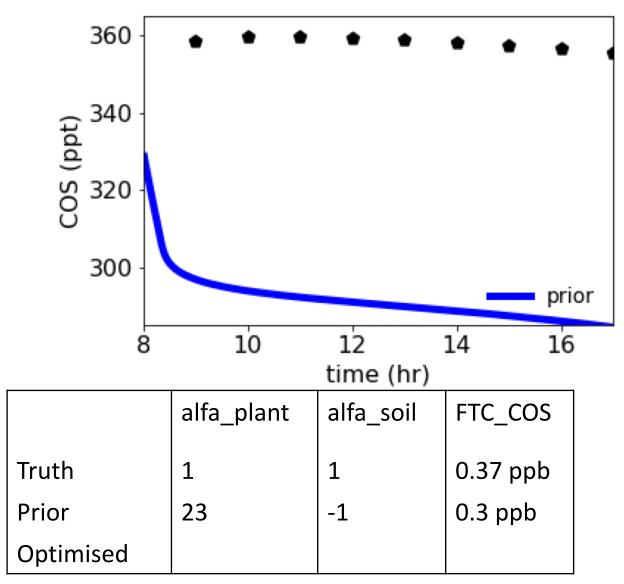
#### Parameters Harvard

Harvard four parameters optimised, two fluxes in cost function

	alfa_plant	alfa_soil	FTC_COS (ppb)	FTC_CO2_scale
Prior	0.8	0.5	0.380	1 (364 ppm)
Optimised	0.822	-4.674	0.361	1.063 (387 ppm)

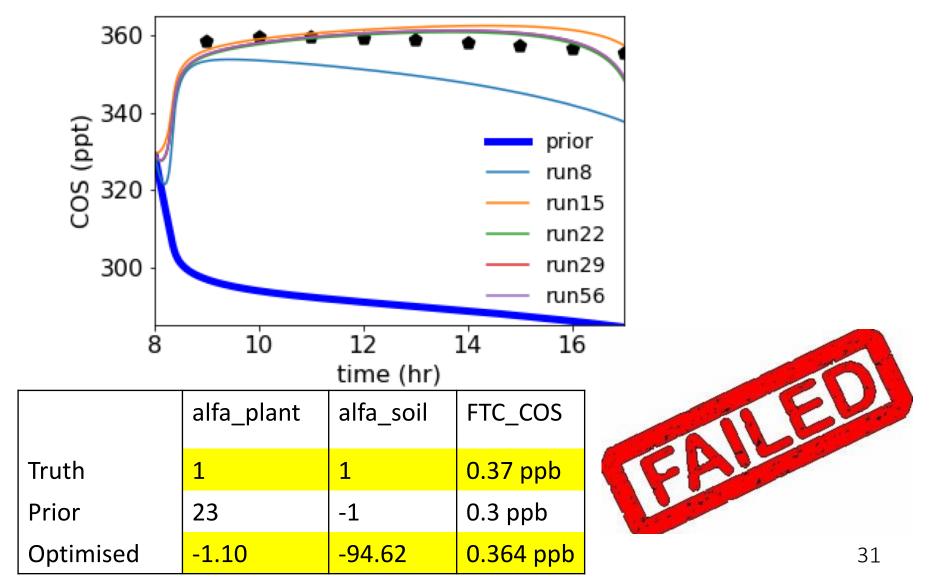
## Optimiser performance test

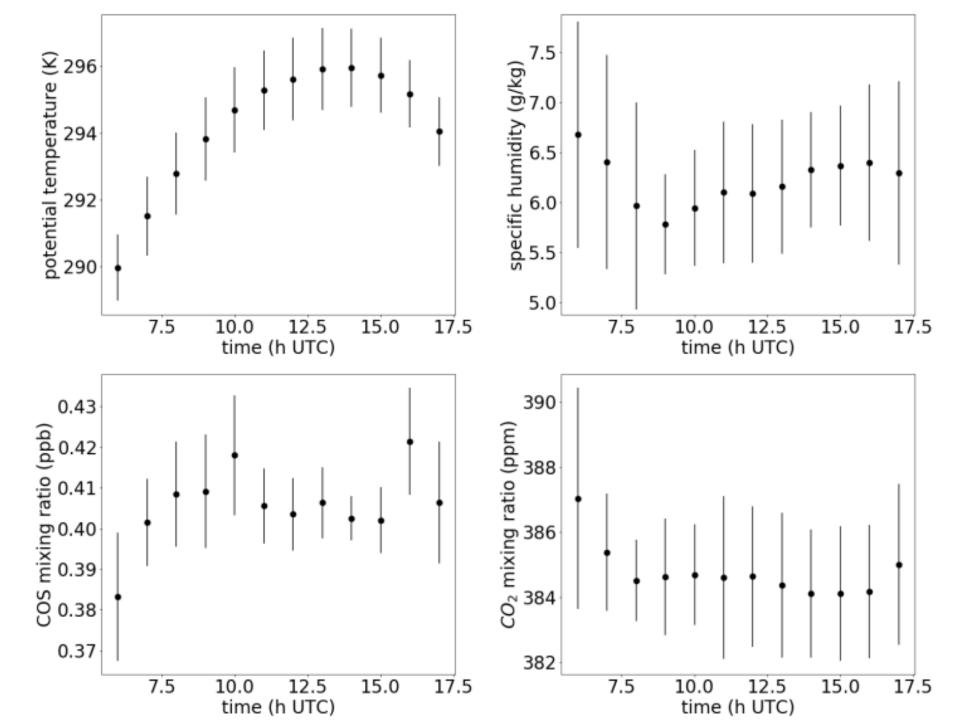
#### Now three parameters



## Optimiser performance test

#### Now three parameters



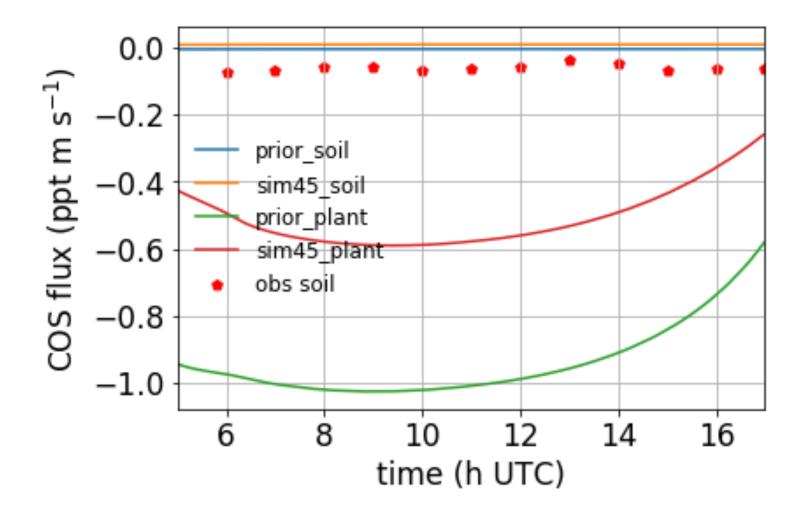


## Current coupling COS-CO<sub>2</sub>

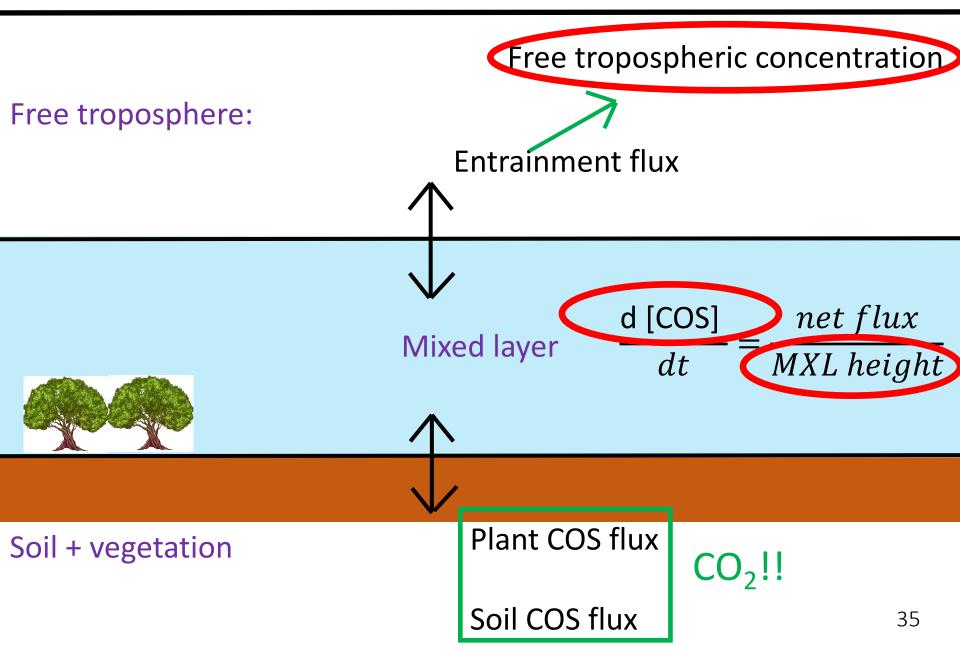
$$gCO_{2}can = gmin * LAI + \frac{f_{str} * a_1 \int_0^{LAI} A_g dL}{\left(C_{ext} - C_{comp}\right) \left(1 + \frac{D_s}{D_*}\right)}$$

$$gsCO_2\_can = gCO_2\_can$$
$$gsCOS\_can = \frac{gsCO_2\_can}{1.21}$$
$$gCOS\_can = \frac{1}{\left(\frac{1}{gsCOS\_can} + \frac{1}{giCOS}\right)}$$

## Soil flux Hyytiala



## Potential of the framework



# Vegetation uptake

COS uptake and photosynthesis coupled to stomatal conductance -> crucial link between COS and photosynthesis



COS in leaves destroyed by enzymes, with limited backward diffusion

CO<sub>2</sub> in leaves often assimilated, but backward diffusion can be significant

