



Big Root approximation of site-scale vegetation water uptake

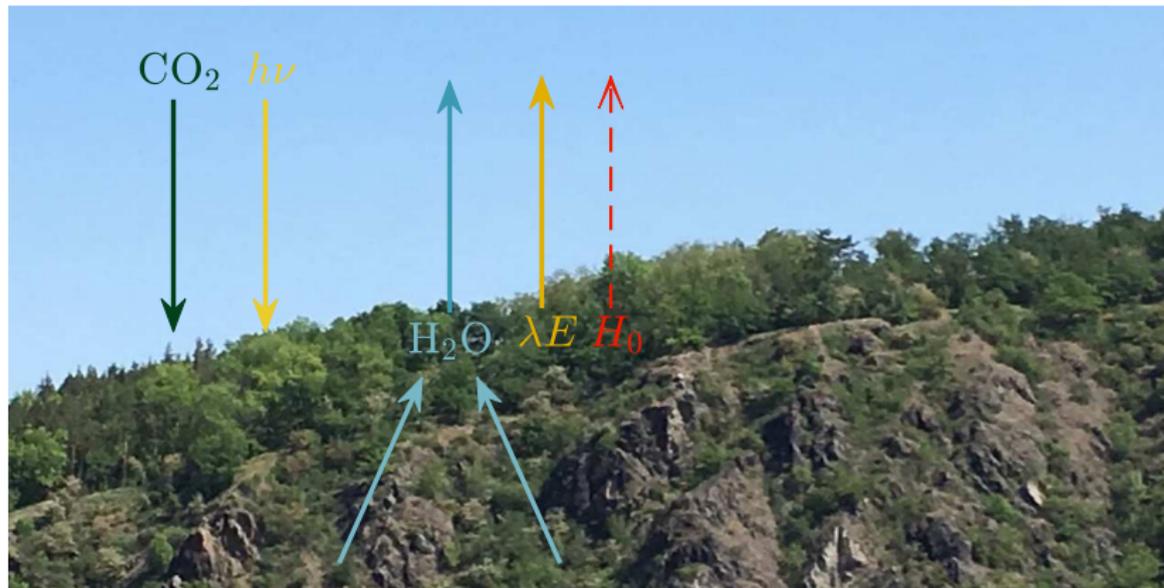
Surfex workshop, Mar. 18, 2019

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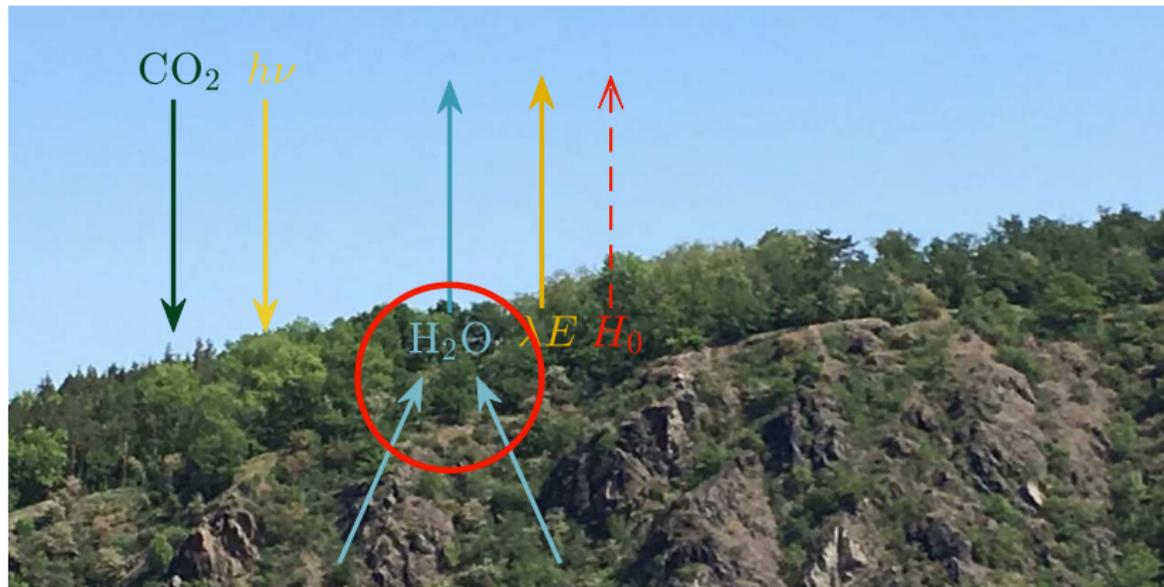
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Vegetation as a Land Surface Feature



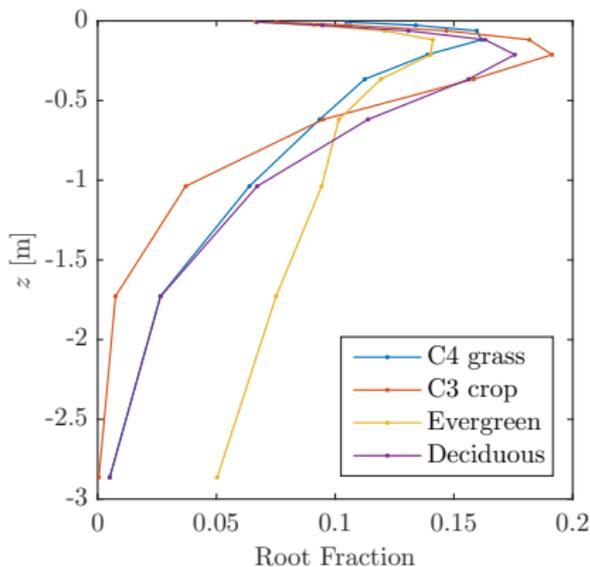
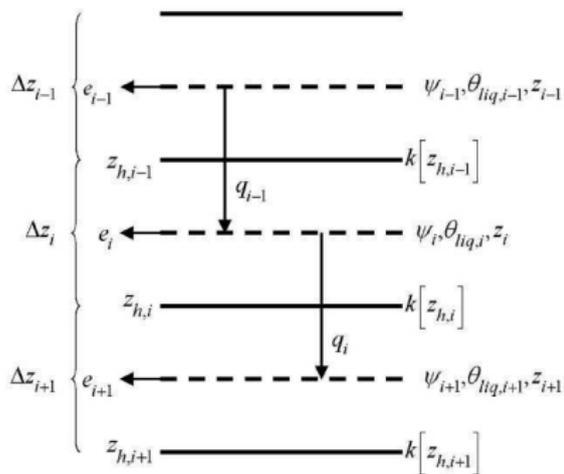
Canyon of Vltava river at Sedlec

Vegetation as a Land Surface Feature



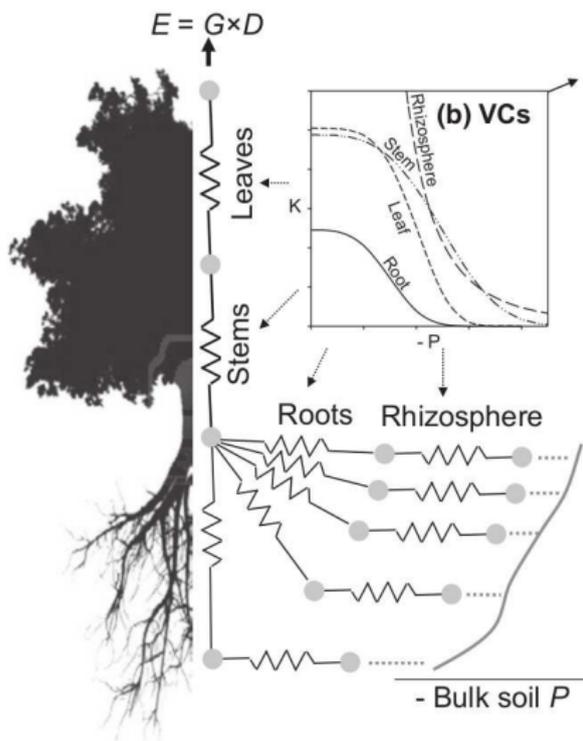
Canyon of Vltava river at Sedlec

Terrestrial models: soil water and root representation



$$\text{stress factor } \beta = \frac{\psi - \psi_{min}}{\psi_{max} - \psi_{min}}$$

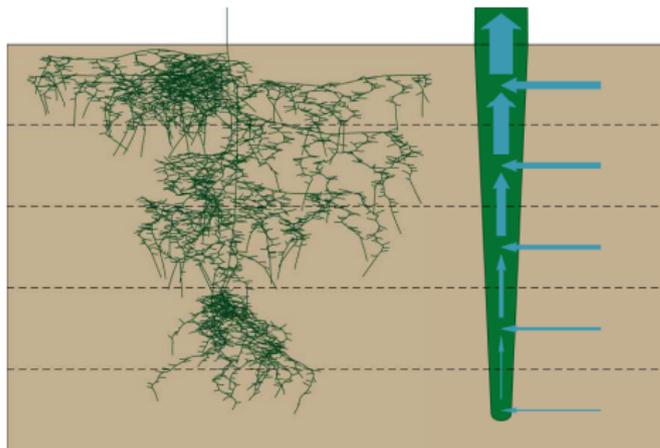
Next generation



(Sperry *et al.*, 2016, *New Phytologist*)

- ▶ Moisture gradient in soil, plant, atmosphere.
- ▶ Ohm's law analogue.
- ▶ All root resistances assumed in parallel.
- ▶ Misrepresentation of root system architecture.
- ▶ Known problems, e.g. redistributes water too freely (Kennedy *et al.*, 2019)

Effects of Root System Architecture

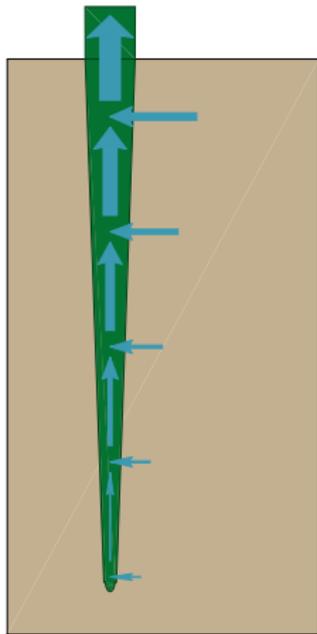


- ▶ Flows accumulate upward.
- ▶ The more inflow in layer i , the greater the difference in water potential across layers $i \pm 1$.
- ▶ RSA determines ratio between uptake and potential gradient dissipation.

$$Q_{plant}^{i \Rightarrow i-1} - Q_{plant}^{i+1 \Rightarrow i} = Q_{soil \Rightarrow plant}^i$$

$$\frac{\Delta^2 \bar{\psi}_x}{\Delta z^2} = \frac{\bar{K}_r^{soil \Rightarrow plant}}{\bar{K}_z^{plant}} (\bar{\psi}_x - \psi_s)$$

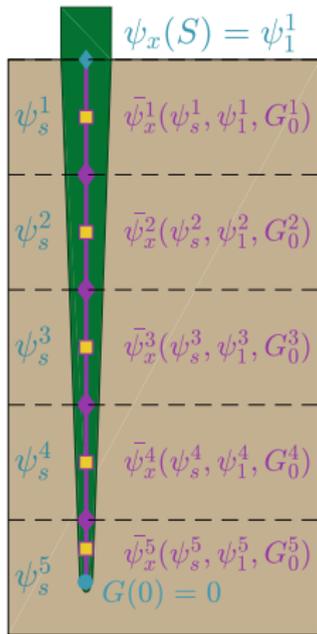
Governing Equation



$$\frac{d^2\psi_x}{ds^2} = \frac{k_r}{K_x}(\psi_x - \psi_s)$$

- ▶ ψ_x (Pa) ‘xylem’ water potential
- ▶ s (m) length along root
(0 at tip, S at base)
- ▶ ψ_s (Pa) ‘soil’ water potential
- ▶ k_r ($\text{m}^2 \text{s}^{-1} \text{Pa}^{-1}$) soil-root ‘radial’
conductance in cross-section
- ▶ K_x ($\text{m}^4 \text{s}^{-1} \text{Pa}^{-1}$) xylem ‘axial’
conductance in cross-section

Segment mean water potential



Can find solutions that yield mean water potential in root segments ($\bar{\psi}_x^i$):

$$\bar{\psi}_x^i = \frac{\int_0^{S^i} \psi_x(s) ds}{S^i}$$

Layer water uptake from Darcian expression using $\bar{\psi}_x$:

$$Q_R^i = -k_r^i S^i (\bar{\psi}_x^i - \psi_s^i)$$

Possible equations for $\bar{\psi}_x^i$

Table: List of variables

$$\bar{\psi}_x^i = \psi_s^i + (G_1^i - G_0^i)/\beta_2^i \quad (1a)$$

$$\bar{\psi}_x^i = c_1^i \psi_0^i + (1 - c_1^i) \psi_s^i - c_2^i G_0^i \quad (1b)$$

$$\bar{\psi}_x^i = c_3^i \psi_0^i + (1 - c_3^i) \psi_s^i - c_4^i G_1^i \quad (1c)$$

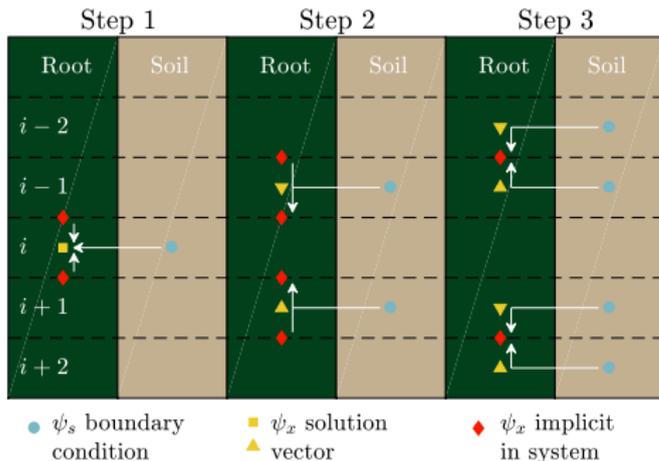
$$\bar{\psi}_x^i = c_1^i \psi_1^i + (1 - c_1^i) \psi_s^i + c_2^i G_1^i \quad (1d)$$

$$\bar{\psi}_x^i = c_3^i \psi_1^i + (1 - c_3^i) \psi_s^i + c_4^i G_0^i \quad (1e)$$

$$\bar{\psi}_x^i = c_5^i (\psi_1^i + \psi_0^i) + (1 - 2c_5^i) \psi_s^i \quad (1f)$$

Symbol	Expression
α^i	$\sqrt{k_r^i/K_x^i}$
β^i	$\alpha^i S^i$
β_2^i	$(\alpha^i)^2 S^i$
c_1^i	$\sinh(\beta^i)/\beta^i$
c_2^i	$(1 - \cosh(\beta^i))/\beta_2^i$
c_3^i	$\tanh(\beta^i)/\beta^i$
c_4^i	$(\operatorname{sech}(\beta^i) - 1)/\beta_2^i$
c_5^i	$\tanh(\beta^i/2)/\beta^i$

Analytical equation for single root $\bar{\psi}_x^i$

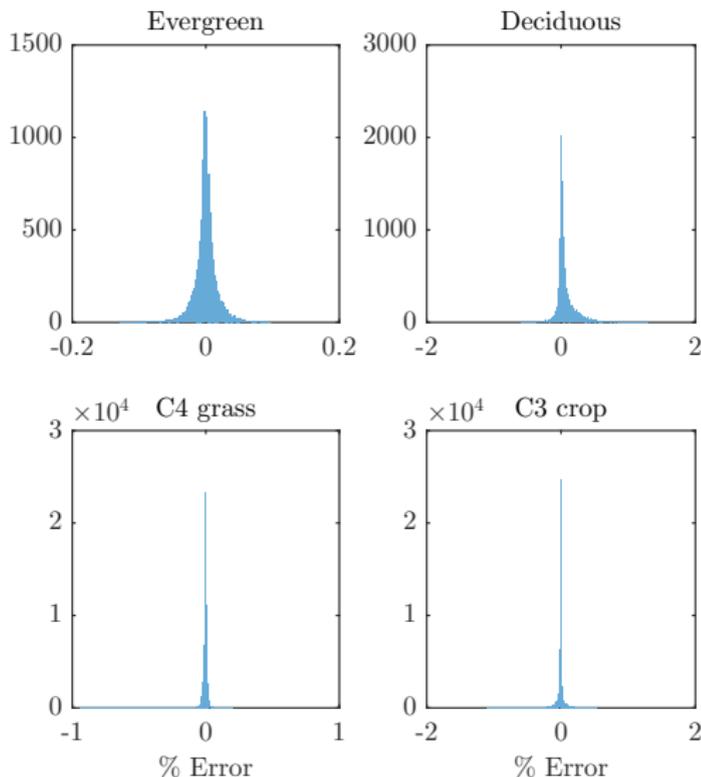
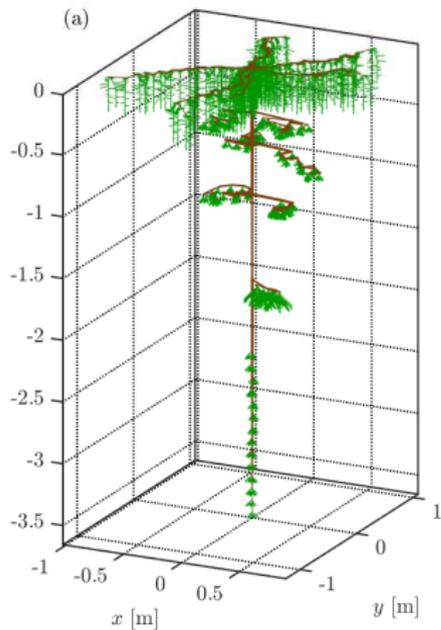


Derivation for layer i :

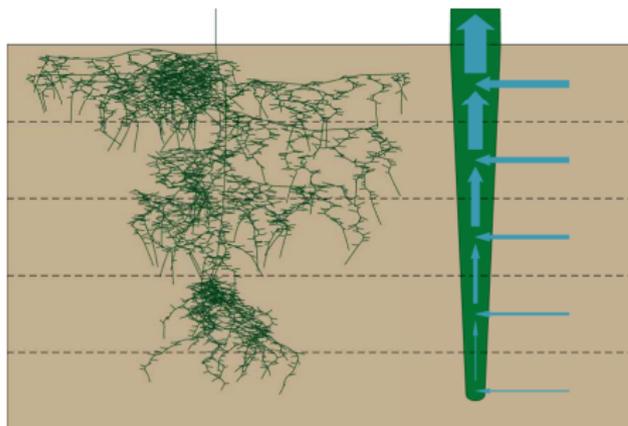
1. start from eq. 1f,
2. substitute for $\psi_{0/1}^i$ from eq. 1f,
3. substitute for $\psi_{0/1}^{i\pm 1}$ from combined eqs. 1b and 1d.

$$\bar{\psi}_x^i = \sum_{j=i-2}^{i+2} a^j \bar{\psi}_x^j + \sum_{k=i-2}^{i+2} b^k \psi_s^k, (j \neq i)$$

RSA Stencil: accurate $\bar{\psi}_x$ & Q_R predictions



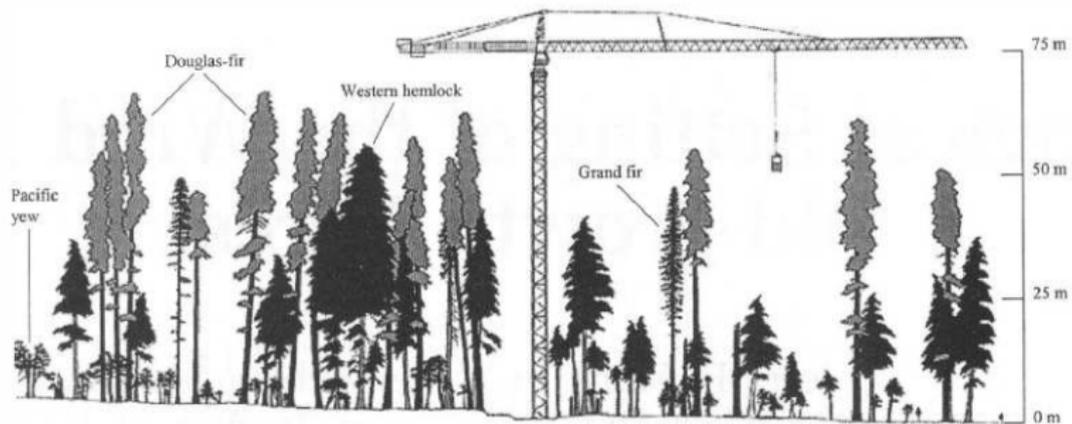
Big Root Model:



Use single-root equations to represent non-trivial RSA

- ▶ Lowers model skill as compared to ‘unconstrained’ RSA-Stencil fit.
- ▶ Provides a clear physical basis:
 - ▶ helps inverse model convergence
 - ▶ more easily interpretable parameters ($k_r S$, K_x/S)
 - ▶ greater consistency in prediction beyond calibration data

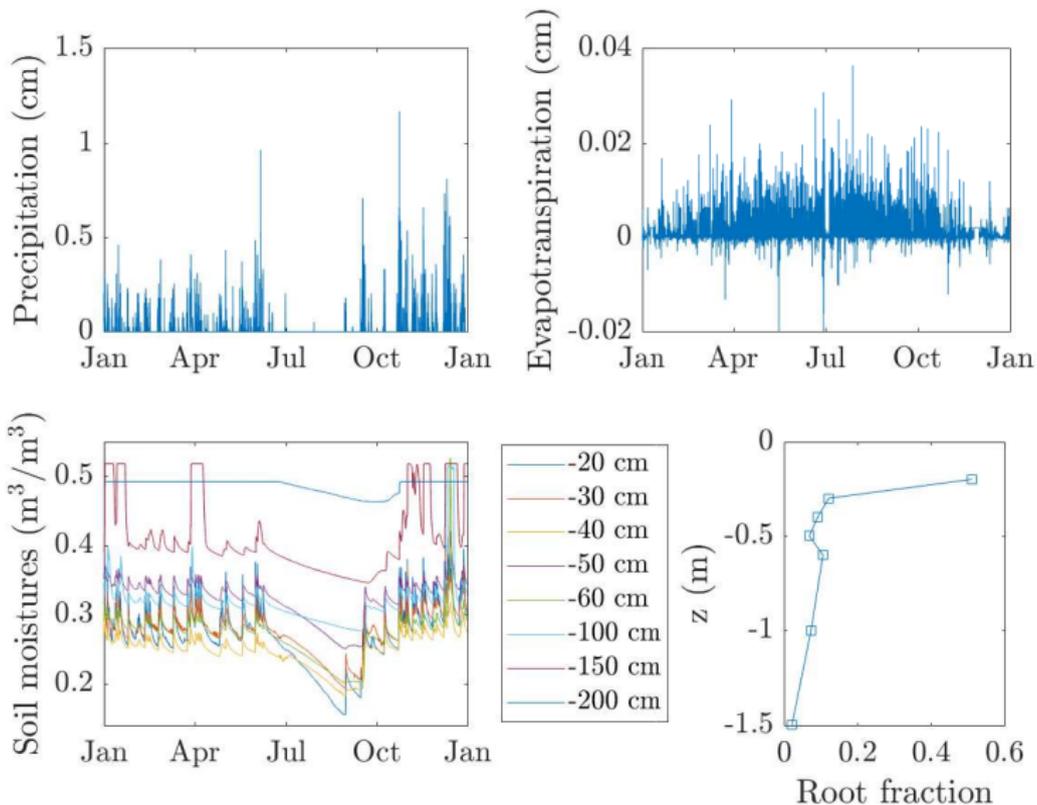
Big Root case study: prediction of field data from Wind River Crane



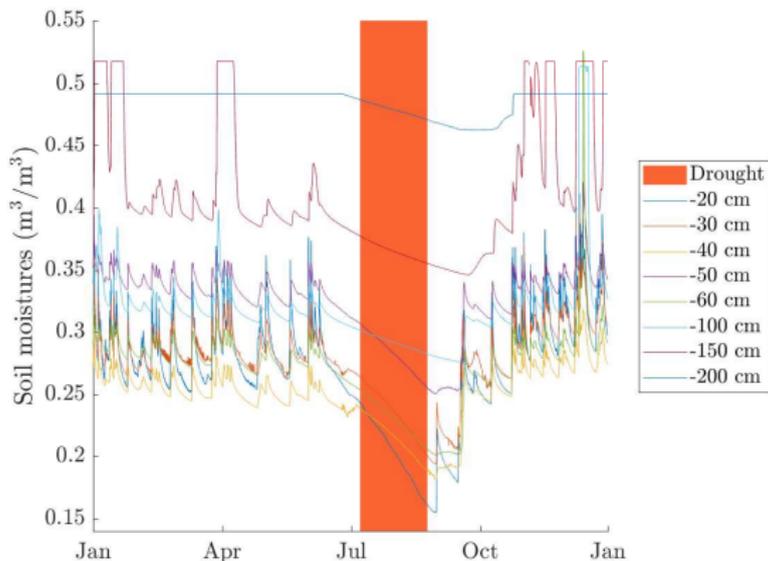
Shaw et al. (2004) *Ecosystems*

(Extensively studied old growth douglas-fir / hemlock site)

2010 Data Overview



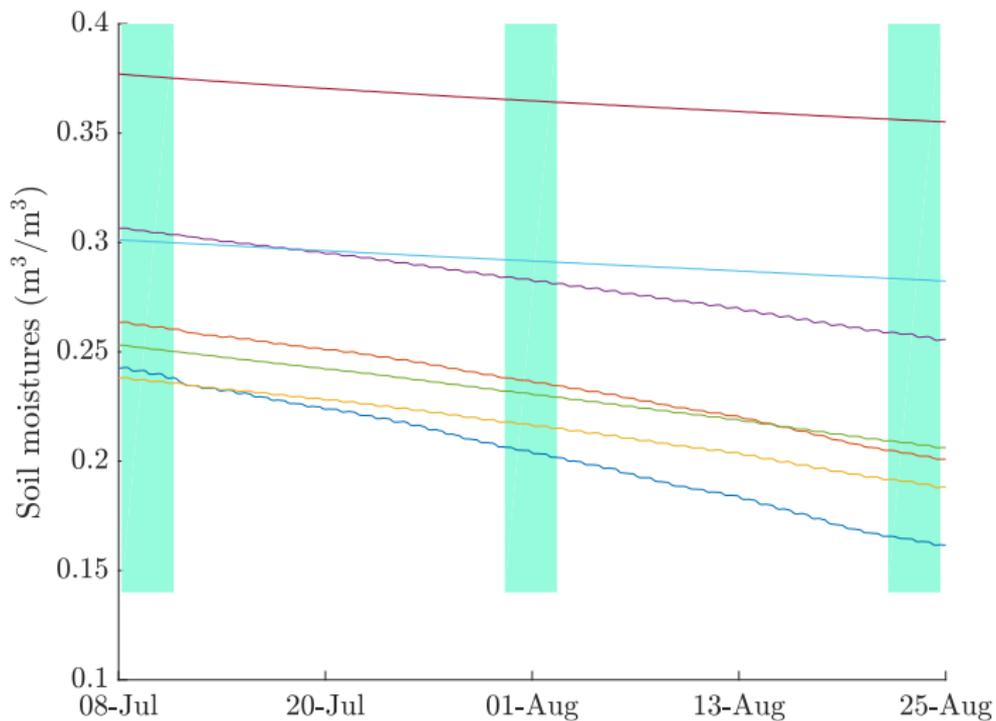
2010 Drought



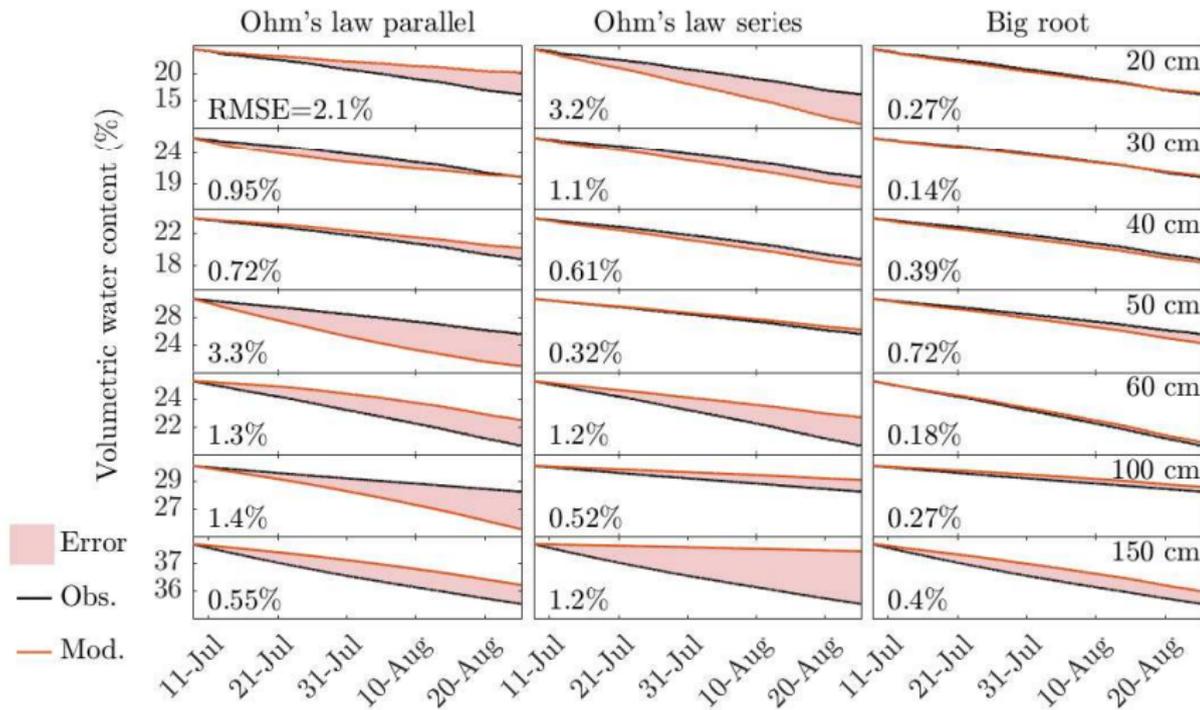
Assumptions:

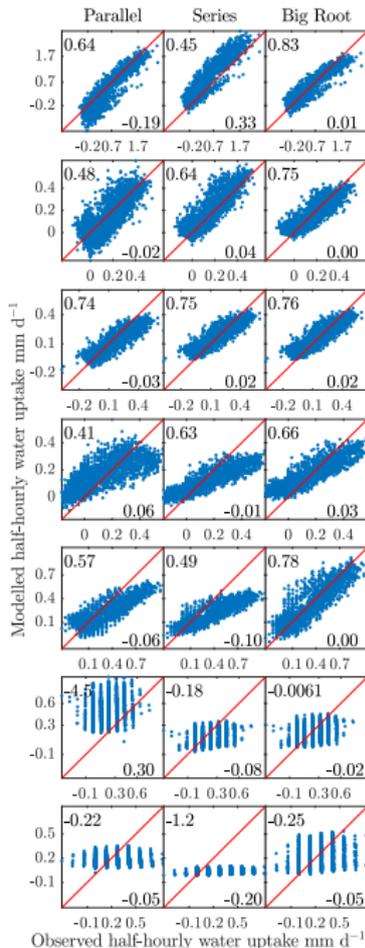
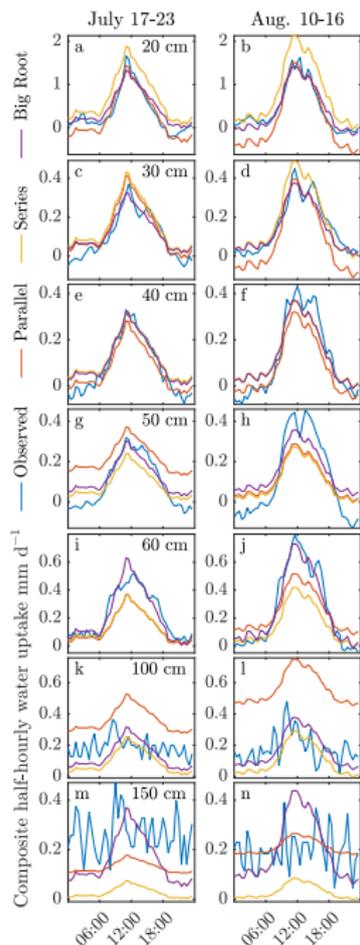
- ▶ all water movement is through plant
- ▶ all ET is transpiration
- ▶ published $\psi_s - \theta$ relations
- ▶ published basal area

Big root calibration data subset



Soil Moisture Predictions



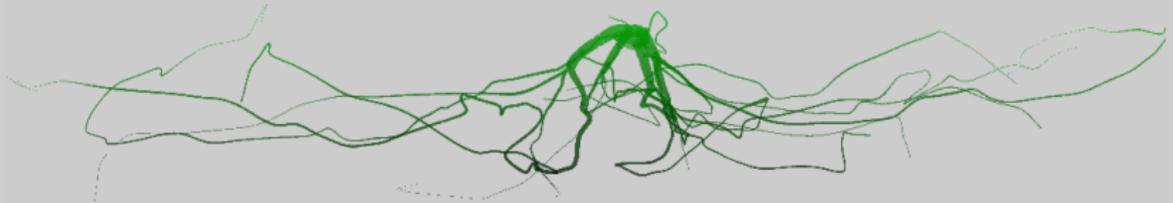


Water Uptake Predictions

- ▶ Ohm's law parallel: overly redistributes dry bias in wet wet bias in dry
- ▶ Ohm's law series: shallow dry bias deep wet bias
- ▶ Big root: more flexible RSA no systematic bias

Conclusions

- ▶ RSA imposed by Ohm's law analogue models leads to biases.
- ▶ Big root model includes interactions between layers, is more flexible at representing RSA.
- ▶ Systematic biases eliminated; errors due to assumption of single effective root; but this also puts predictions on a physical basis, increasing robustness.
- ▶ Next steps (in cooperation with CNRM, Météo France):
 - ▶ implement big root model in SurfEx,
 - ▶ show impact of root scheme on soil moistures and surface fluxes at scale.



Acknowledgements

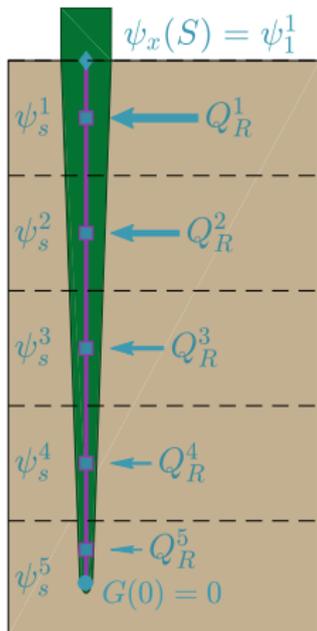
- ▶ Many thanks to Sonia Wharton and the Wind River Crane team!
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Model inversion



- ▶ Cannot uniquely resolve k_r^i , K_x^i , S^i without further constraint.
- ▶ Inversion yields $k_r^i S^i$ and K_x^i / S^i or just stencil a and b , predicting flows (Q_R):

Data	Boundary conditions	Inversion yields:
$\bar{\psi}_x, Q_R$	$\psi_1^1, G_0^n = 0$	$k_r S, K_x / S$
$\bar{\psi}_x, Q_R$	$G_1^1, G_0^n = 0$	$k_r S, K_x / S$
Q_R	$\psi_1^1, G_0^n = 0$	$k_r S, K_x / S^*$
Q_R	$G_1^1, G_0^n = 0$	stencil a, b

* except layer n

