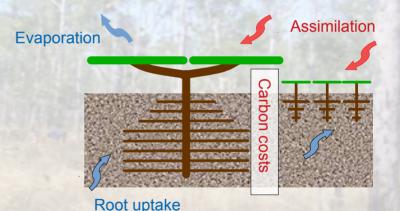


Evaluation of the Vegetation Optimality Model along the North-Australian Tropical Transect using a fully Open Science approach

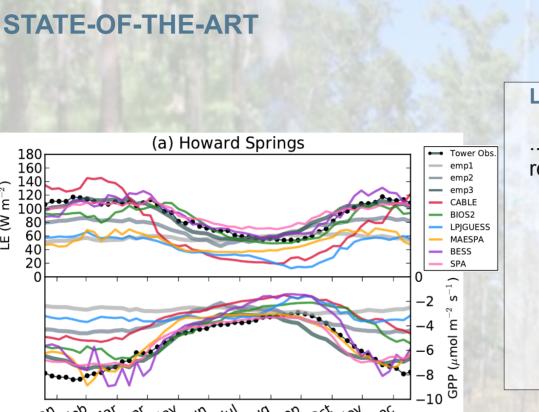
R.C. Nijzink¹, J. Beringer², L.Hutley³, C. Ramakrishnan⁴, R. Roskar⁴, S. Schymanski¹

¹ Luxembourg Institute of Science and Technology, Belvaux, Luxembourg,
 ²University of Western Australia, Crawley, Australia
 ³Charles Darwin University, Darwin, NT, Australia
 ⁴Swiss Data Science Center, Zurich, Switzerland



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Supported by the Luxembourg National Research Fund (FNR) ATTRACT programme (A16/SR/11254288) Luxembourg National Research Fund

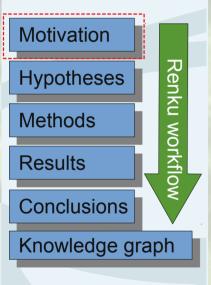


Land Surface Models today:

INSTITUTE OF SCIENCE AND TECHNOLOGY

Luxembourg National **Research** Fund

...produce very different results for unclear reasons



LE (W m⁻²)

13" Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec Whitley et al. (2015): Biogeosciences 13







Land Surface Models today:

...produce very different results for unclear reasons

Tower Obs.

emp1

emp2 emp3

CABLE

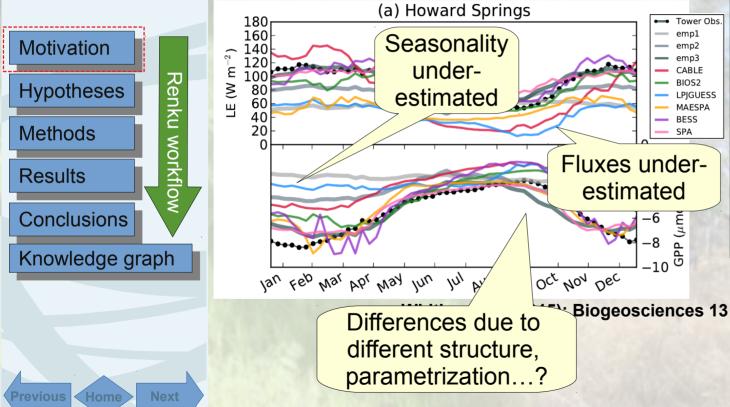
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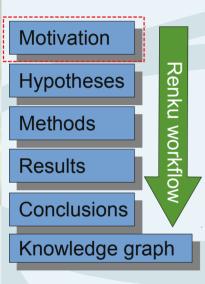




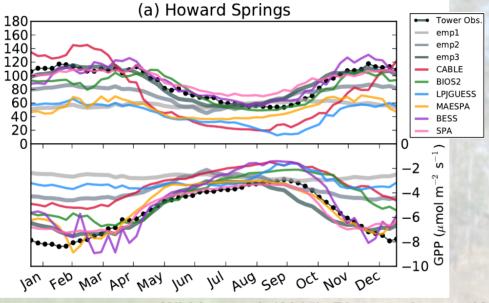
Land Surface Models today:

...produce very different results for unclear reasons

...commonly rely on past observations



LE (W m⁻²)



Whitley et al. (2015): Biogeosciences 13



STATE-OF-THE-ART

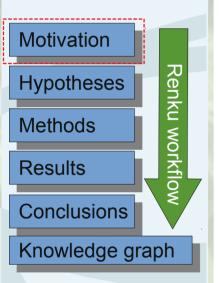




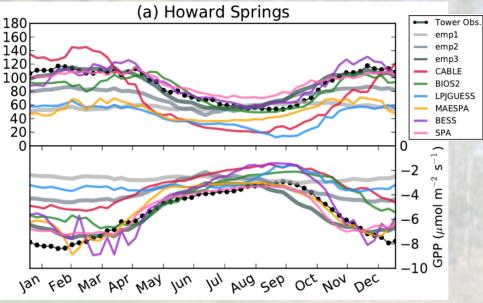
...produce very different results for unclear reasons

...commonly rely on past observations

...cannot simulate full response to change due to prescribed properties



LE (W m



Whitley et al. (2015): Biogeosciences 13



STATE-OF-THE-ART

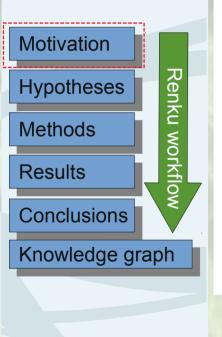




...produce very different results for unclear reasons

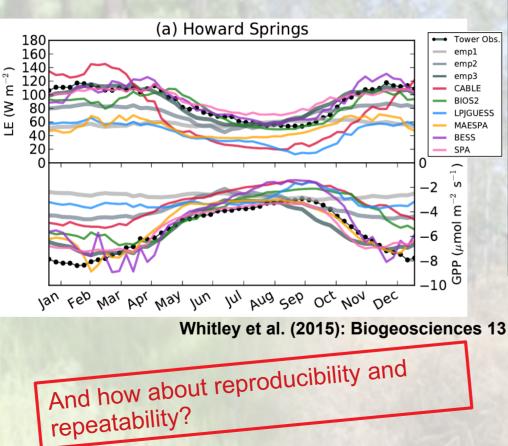
...commonly rely on past observations

...cannot simulate full response to change due to prescribed properties



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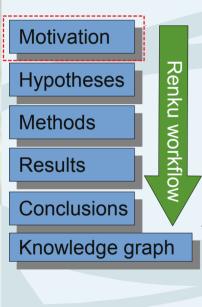
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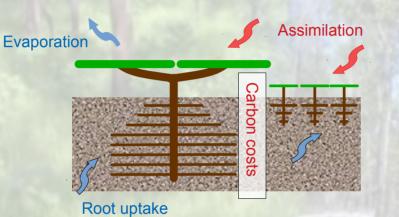
Net Carbon Profit :

Difference carbon uptake and carbon costs



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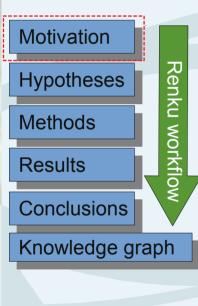


VEGETATION OPTIMALITY



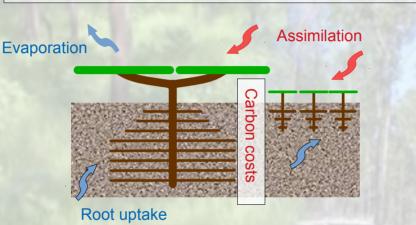
Net Carbon Profit :

Difference carbon uptake and carbon costs



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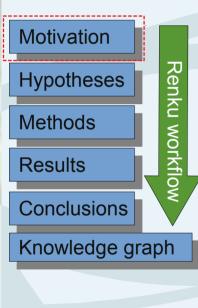
Vegetation Optimality Model Optimizes vegetation properties to maximize NCP

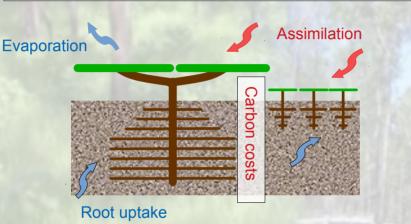




Net Carbon Profit :

Difference carbon uptake and carbon costs

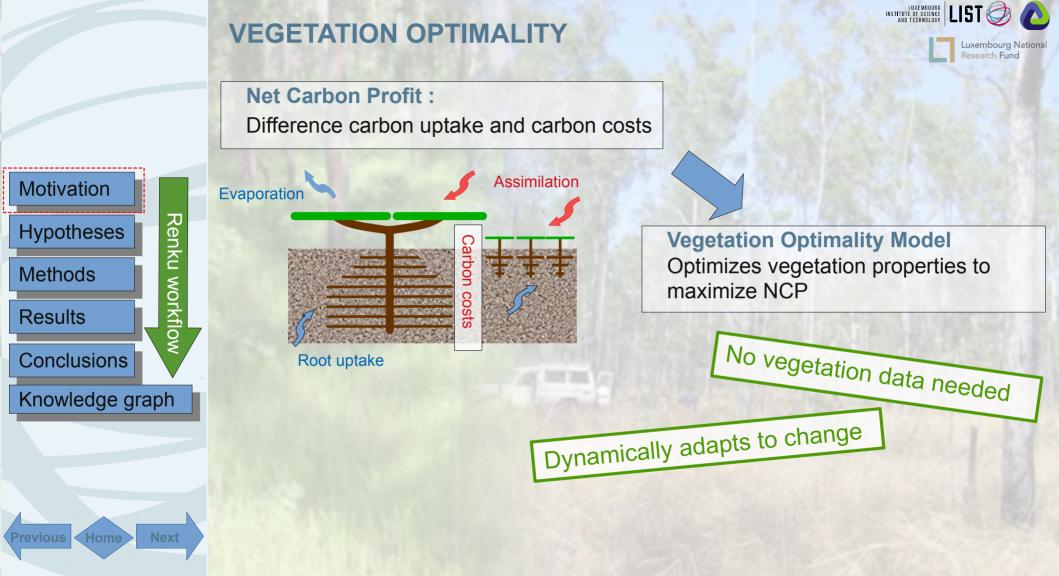


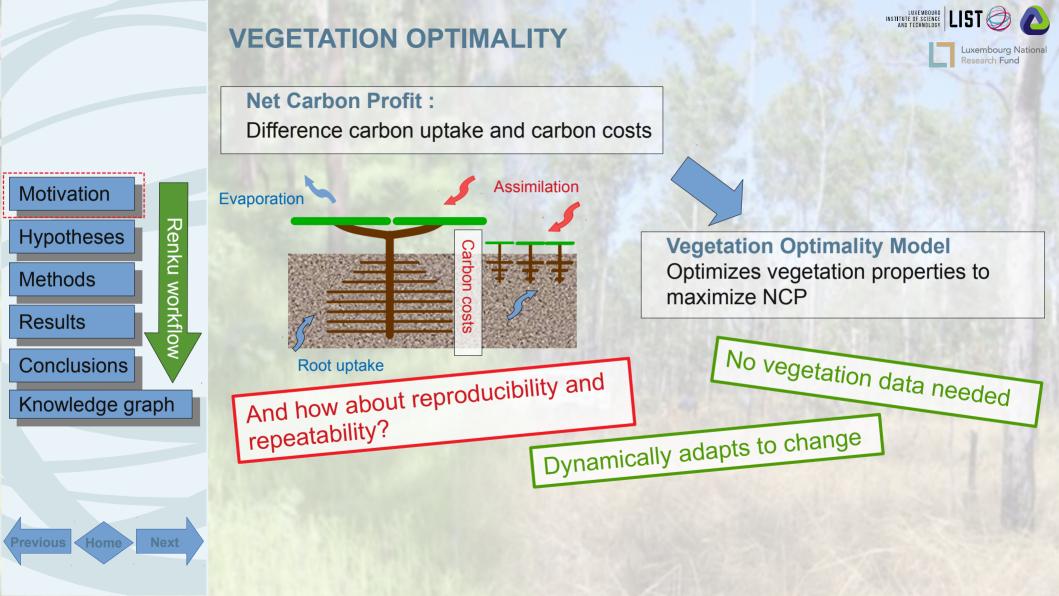


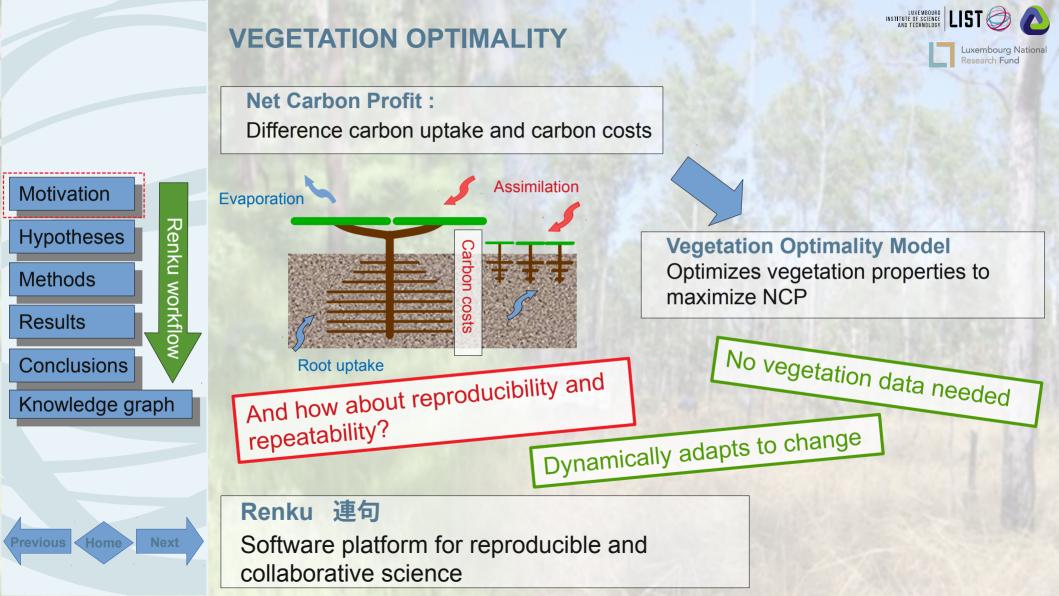
Vegetation Optimality Model Optimizes vegetation properties to maximize NCP

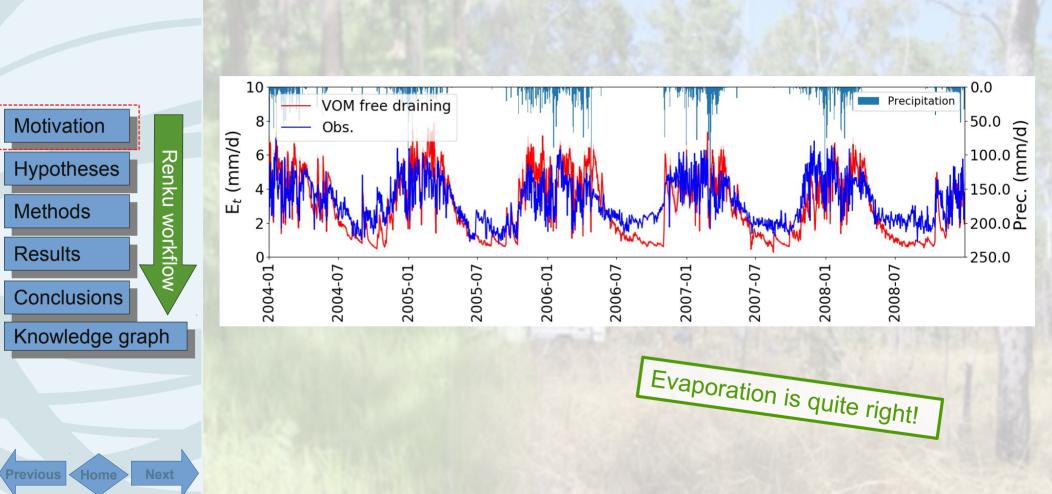






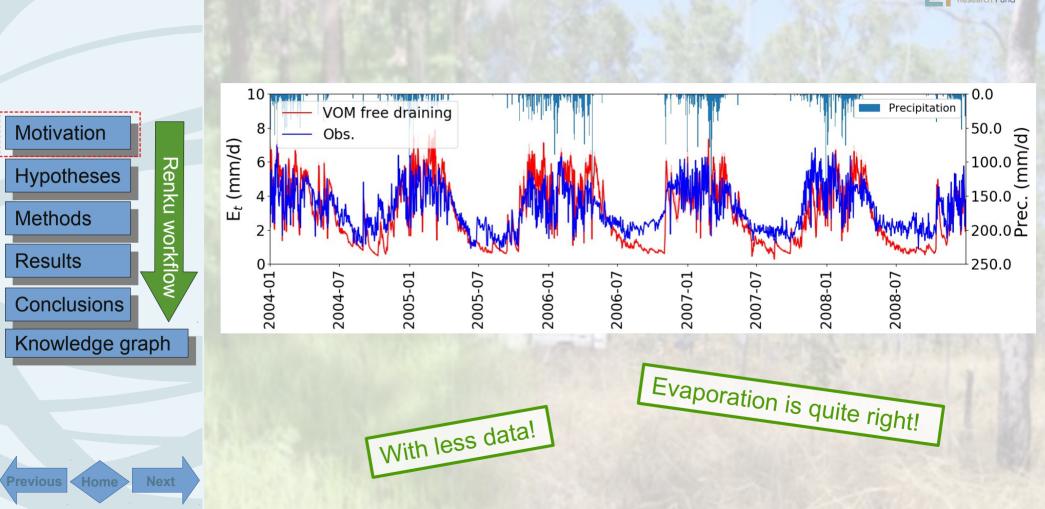






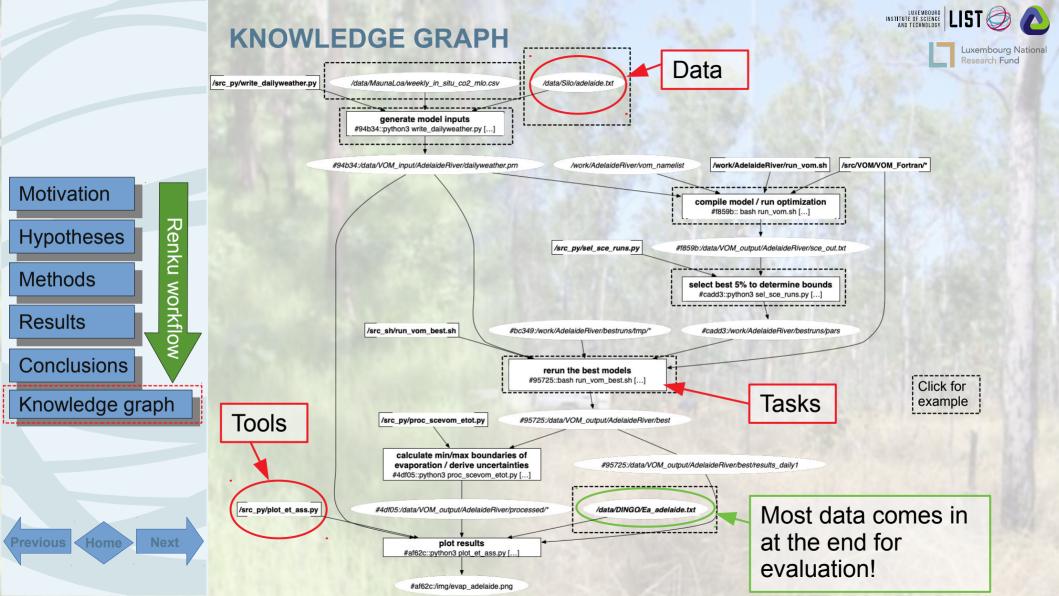
PREVIEW...





PREVIEW...





Evaluation of the Vegetation Optimality Model along the North-Australian Tropical Transect using a fully Open Science approach

R.C. Nijzink¹, J. Beringer², L.Hutley³, R. Roskar⁴, S. Schymanski¹



State-of-the art **Motivation** Renku workflow Vegetation Optimality **Hypotheses Highlights** VOM **Methods** NATT Initial Results Model Comparison Hydrology and carbon costs Vegetation dynamics Results **Conclusions** ¹Luxembourg Institute of Science and Technology, Belvaux, Luxembourg, ² University of Western Australia, Crawley.Australia ³Charles Darwin University, Darwin, NT, Knowledge graph Australia ⁴ Swiss Data Science Center, Zurich, Switzerland

STATE-OF-THE-ART

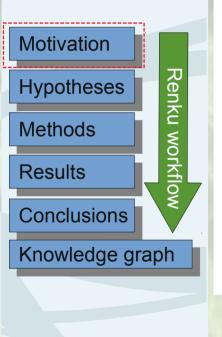




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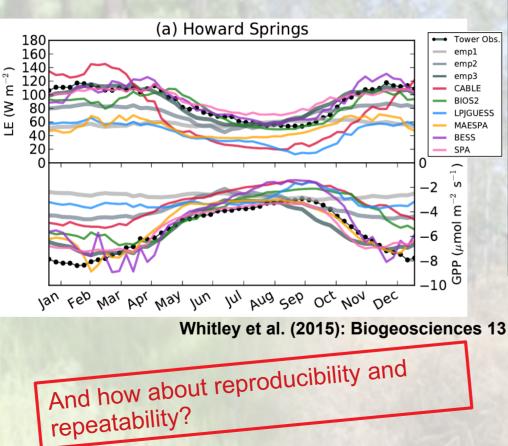
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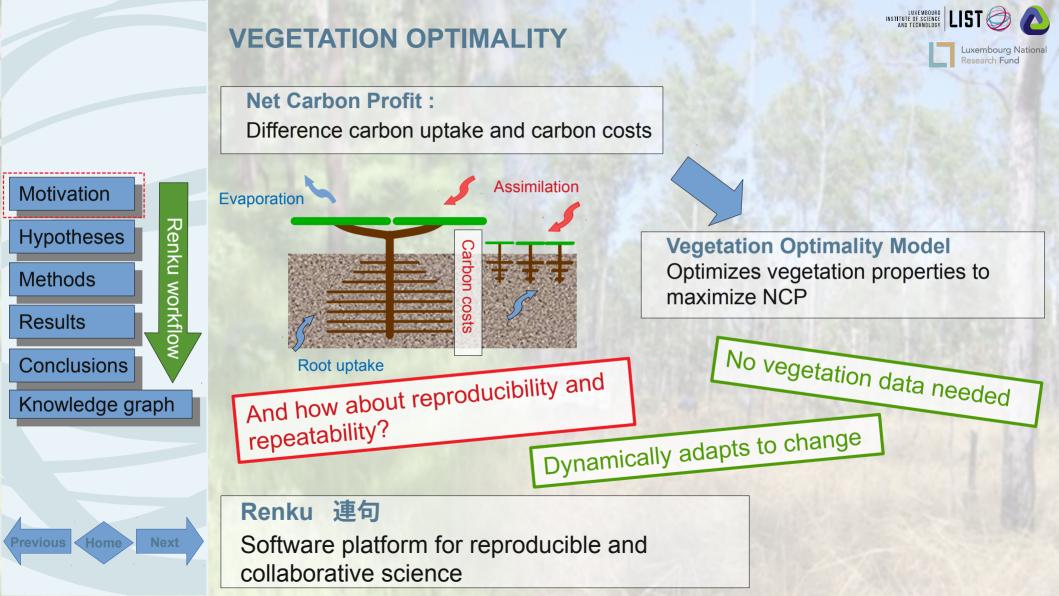
...cannot simulate full response to change due to prescribed properties



Next

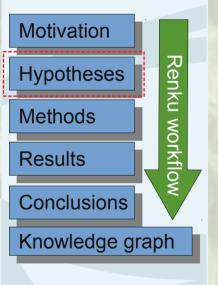
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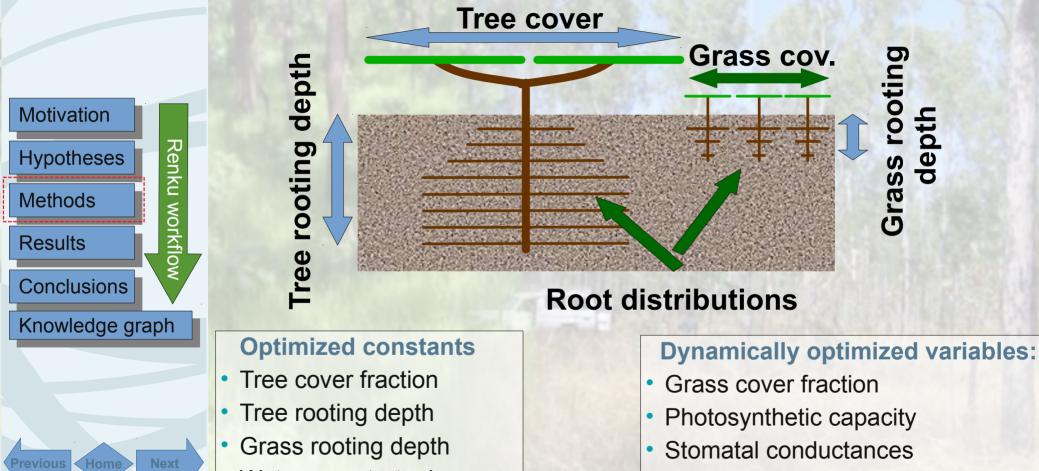
- Observed vegetation dynamics in tropical savanna sites can be explained by the maximization of Net Carbon Profit.
- Optimization of vegetation properties for the Net Carbon Profit leads to reduced data requirements for Land Surface Models
- Hydrological formulation of Land Surface Models matters for flux exchanges



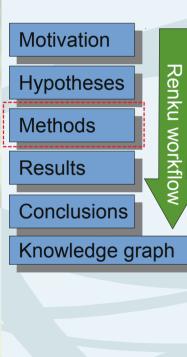
VEGETATION OPTIMALITY MODEL



Fine root surface area



Water use strategies



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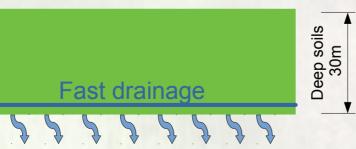
VEGETATION OPTIMALITY MODEL

Getting the hydrology right:

- Groundwater influences root water uptake
- Water balance should be correct
- Modelling for the right reason

Free draining conditions

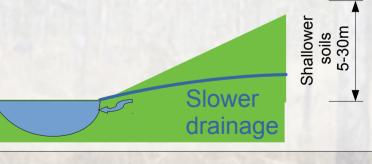
- Conventional approach
- Large unsaturated zone (Zr = 30m)
- No influence of groundwater table

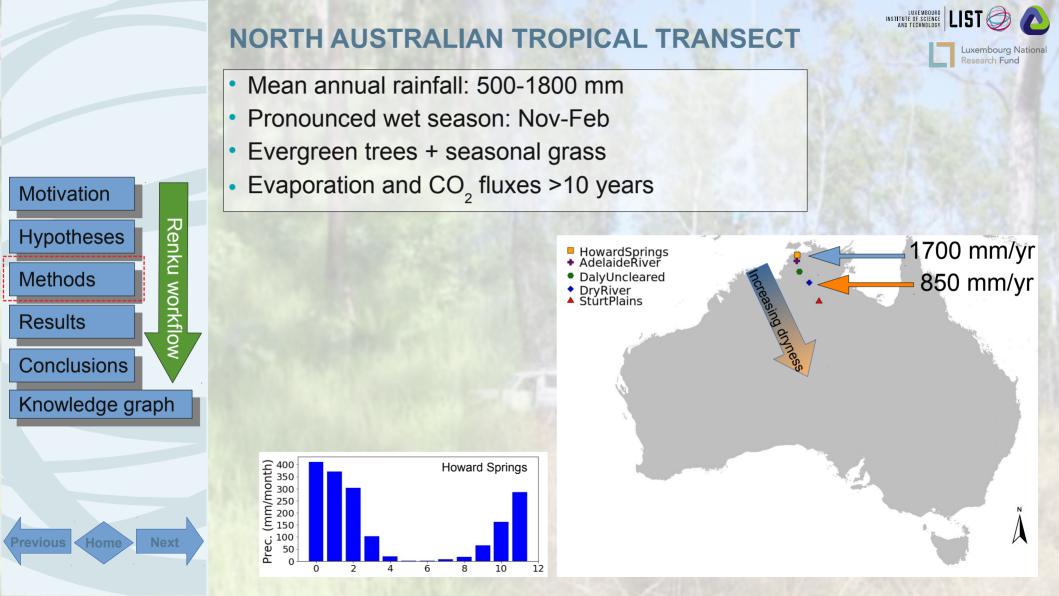




Dynamic water tables

- Unsaturated zone from 5 30m
- Based on max. elevation and stream elevation
- Drainage depends on slopes
- Influence of groundwater tables







RENKU 連句

Renku is an environment for collaborative, reproducible data science

Concepts

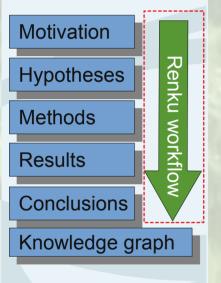
- Tracking of scientific steps to create data lineage, i.e. a knowledge graph
- Updating of out-dated results
- Tool to re-use or re-run analyses
- Sharing of analyses

Features

Renku is based on :

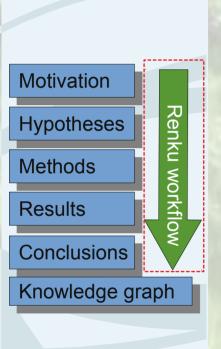
- Gitlab
- JupyterHub
- Kubernetes
- Keycloak
- Common Workflow Language

Go to RENKU website →



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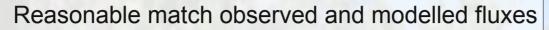


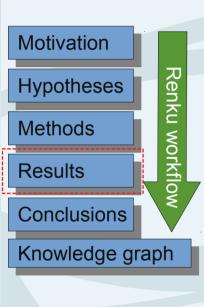
Renku run makes sure the workflow is tracked	Renku run
Renku status shows if all outputs are generated from the most recent input data	Renku status
Renku update re-runs everything to have all outputs based on the most recent inputs	Renku update
<i>Renku log</i> shows how a file is generated. In other words, is shows the knowledge graph.	Renku log Show knowledge graph

Go to workflow of this experiment→

INITIAL RESULTS

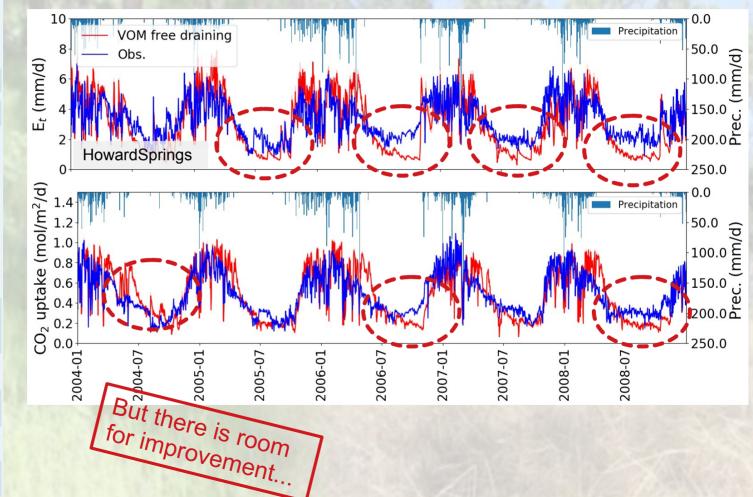






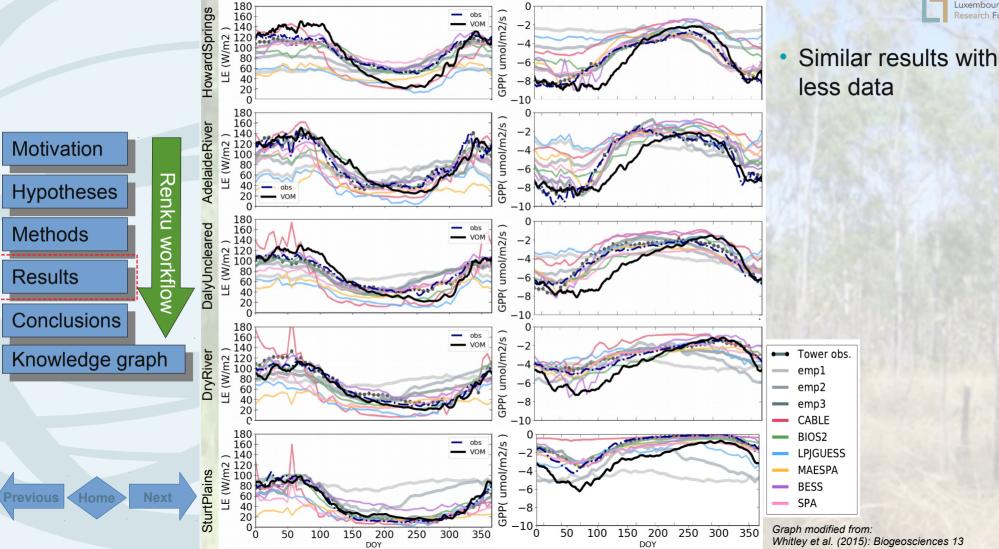
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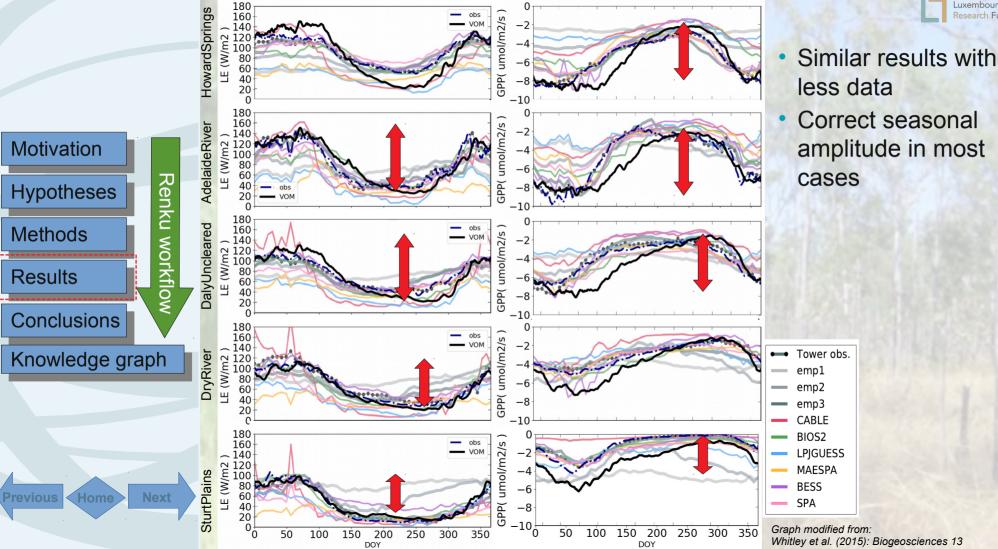
MODEL COMPARISON





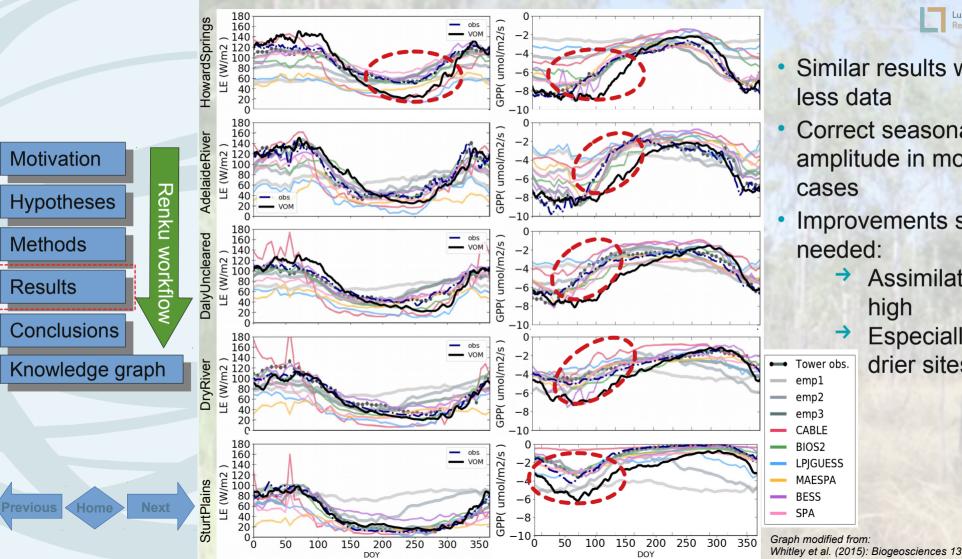
MODEL COMPARISON





Whitley et al. (2015): Biogeosciences 13

MODEL COMPARISON





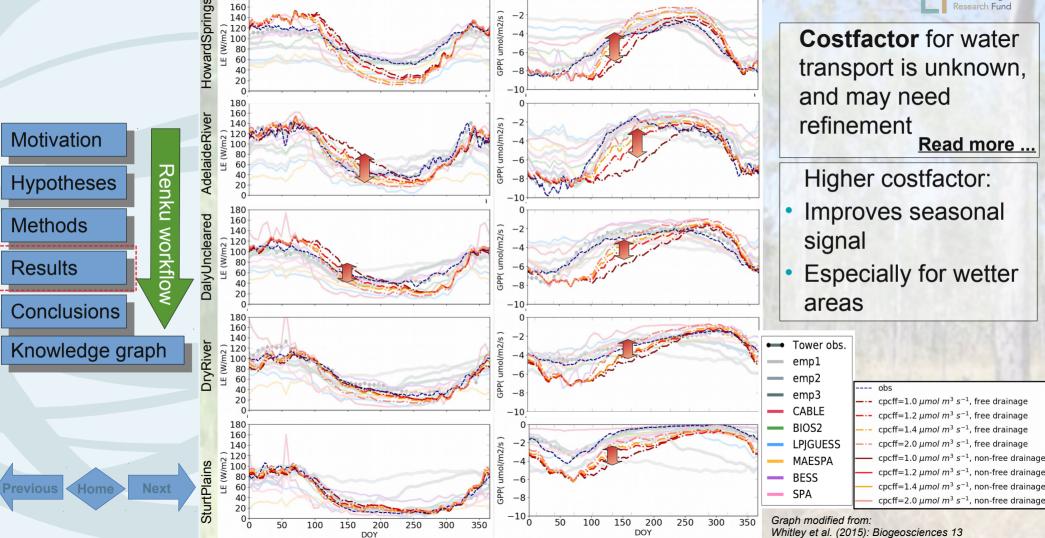
- Similar results with less data
- Correct seasonal amplitude in most cases
- Improvements still needed:
 - \rightarrow Assimilation too high
 - Especially for drier sites

Highlights

MODEL COMPARISON

180





Highlights

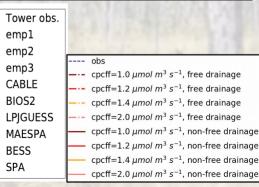
MODEL COMPARISON



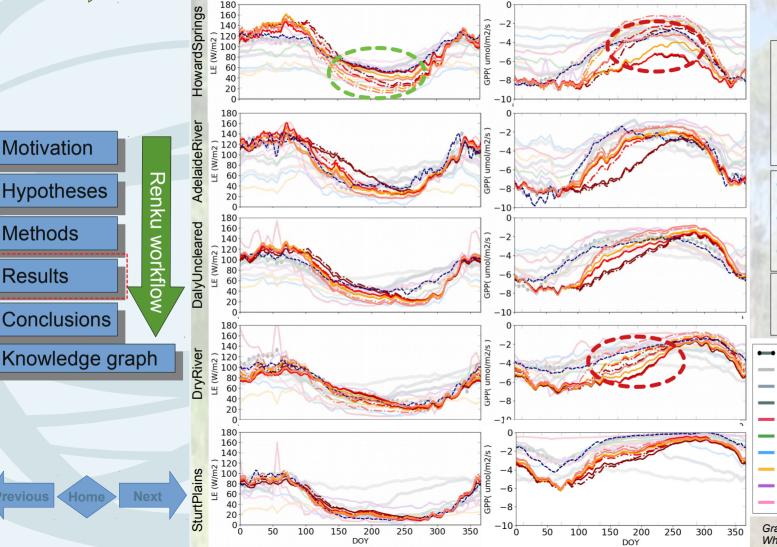
Hydrology is parameterized for free draining and non-free draining conditions

- Some improvements
- Some deteriorations
- Mostly similar results

Uniqueness of place? Robust model?



Graph modified from: Whitley et al. (2015): Biogeosciences 13

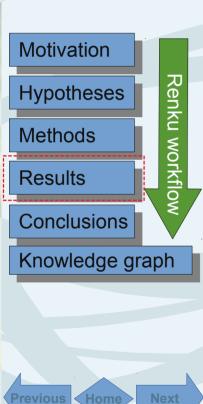


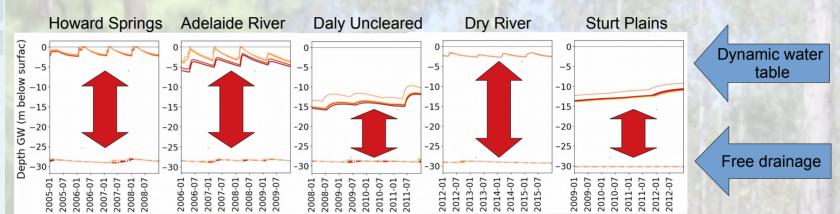
IMPROVING HYDROLOGY AND CARBON COSTS

LUST WOURG AND TECHNOLOGY LLIST OF COLORED LUXEMBOURG National Research Fund

See more..

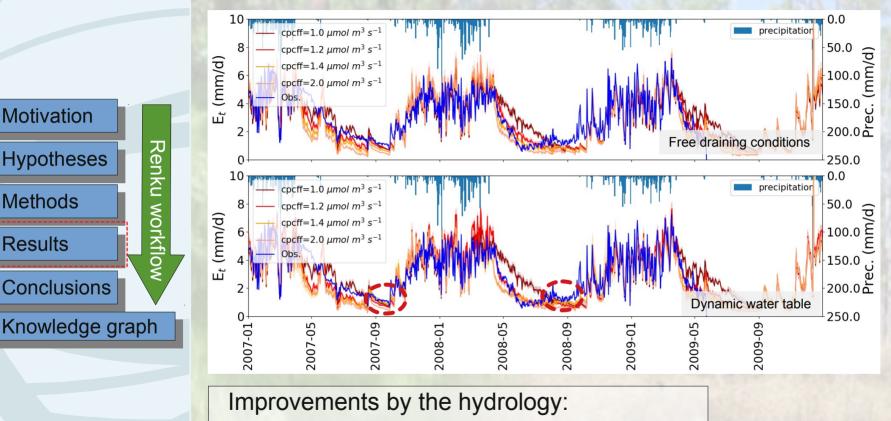
Hydrology differs strongly, but has hardly any influence on fluxes





- --- cpcff=1.0 μ mol m³ s⁻¹, free drainage
- --- cpcff=1.2 μ mol m³ s⁻¹, free drainage
- --- cpcff=1.4 μ mol m³ s⁻¹, free drainage
- --- cpcff=2.0 μ mol m³ s⁻¹, free drainage
- cpcff=1.0 μ mol m³ s⁻¹, dynamic water table
- cpcff=1.2 μ mol m³ s⁻¹, dynamic water table
- cpcff=1.4 $\mu mol~m^3~s^{-1}$, dynamic water table
 - cpcff=2.0 μ mol m³ s⁻¹, dynamic water table

IMPROVING HYDROLOGY AND CARBON COSTS



Only small improvements at end of dry season

Motivation

Hypotheses

Conclusions

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Methods

Results

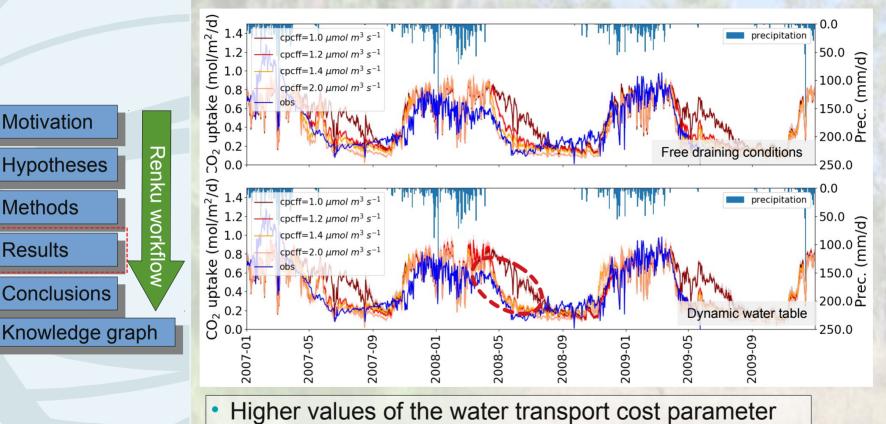
See more ... Howard Springs AdelaideRiver DalyUncleared DryRiver **SturtPlains**

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IMPROVING HYDROLOGY AND CARBON COSTS





See more ... Howard Springs AdelaideRiver DalyUncleared DryRiver **SturtPlains**



Motivation

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Conclusions

Methods

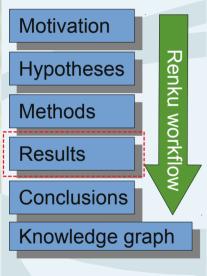
Results

Only small differences for dynamic water tables

improve assimilation

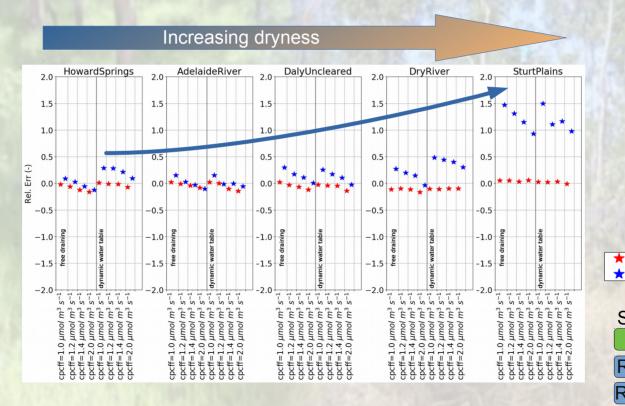
PERFORMANCES

- Assimilation increasingly over-estimated for drier areas
- Evaporation still okay



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See more ... Rel. Err. Annual Means Rel. Err. Mean Dry Season Rel. Err. Mean Wet Season Kling-Gupta Efficiency

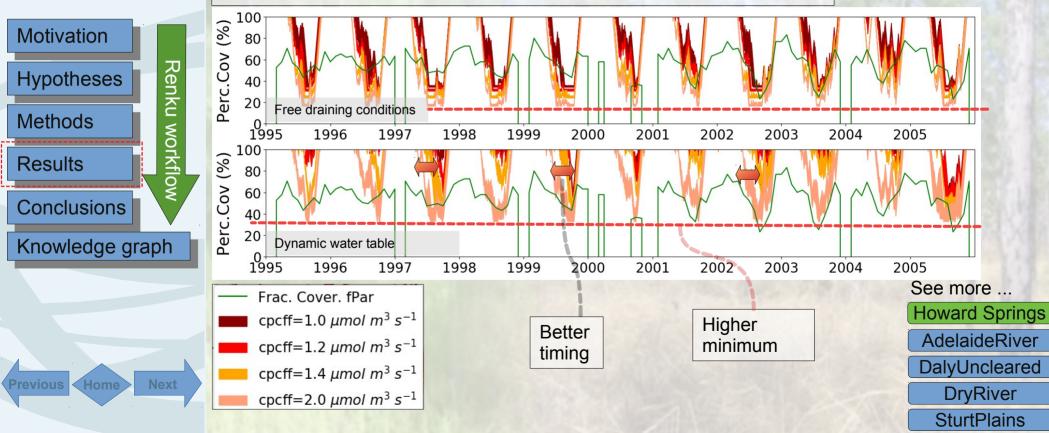
evaporation

assimilation



VEGETATION DYNAMICS

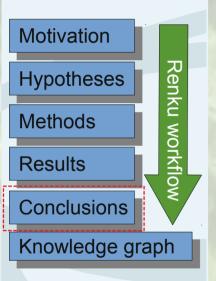
- Temporal signal largely reproduced
- Timing improves for higher cost factor
- Higher minimum cover for non-freely draining conditions







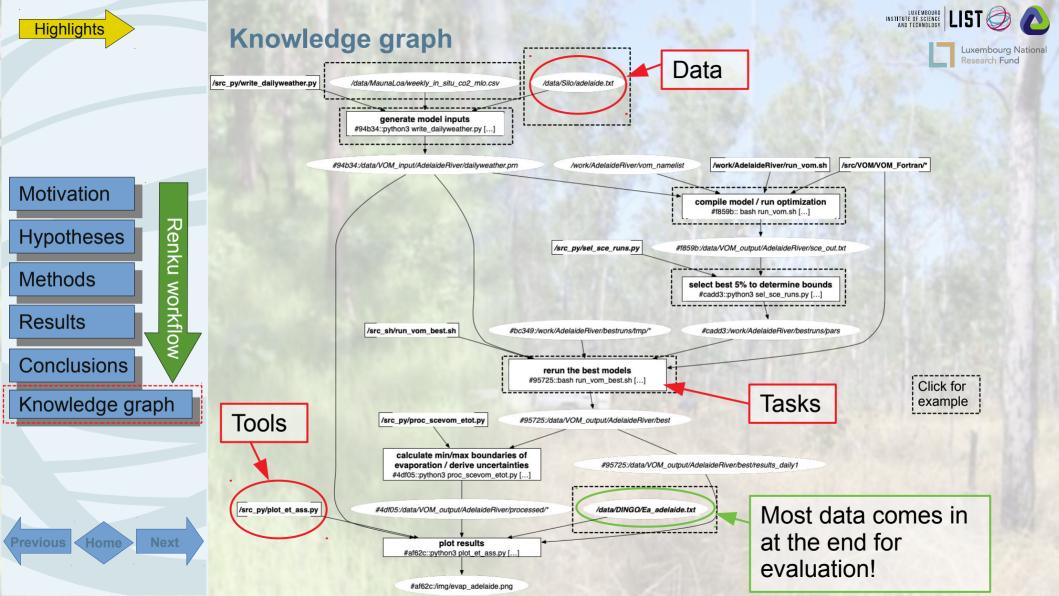


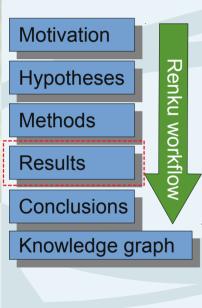


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- Optimizing for the Net Carbon Profit leads to similar vegetation dynamics as observed
- Similar performances as conventional models, with less data
- Not a clear influence of the hydrological formulation
- Cost factor for water transport needs to be refined
- Reproducible science with Renku!





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CARBON COSTS



- **Root respiration** is a function of respiration rate (c_{Rr} , mol s⁻¹ m⁻³), fine root radius (r_r , m), root surface area per unit ground area (S_{Ar} , $m^2 m^{-2}$):
 - $R_r = c_{Rr} \left(\frac{r_r}{2} S_{Ar} \right)$
- Leaf area costs are a function of vegetated fraction (M_A, -), clumped leaf area index (2.5, -), average carbon investment (0.22 μmol s⁻¹ m⁻²):

$$R_f = 2.5 \times 0.22 \,\mu mols^{-1} m^{-2} M_A$$

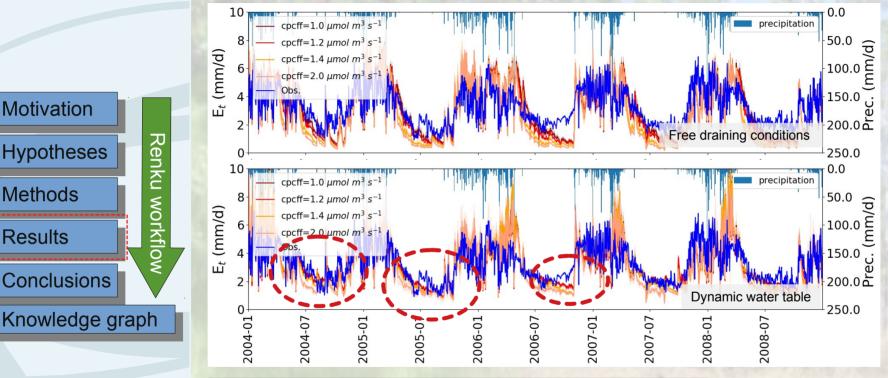
Water transport costs are a function of rooting depth (y_r), vegetated fraction (M_A, -) and a cost factor (cpcff, mol s⁻¹ m⁻³):

 $R_v = cpcff * M_A y_r$

The cost factor cpcff is rather unknown, and may need refinement.



HYDROLOGY AND CARBON COSTS



Improvements by the hydrology:

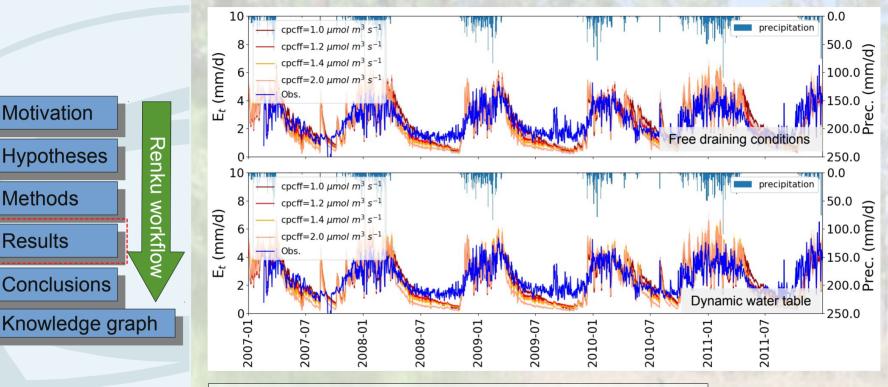
- Dynamic groundwater tables improve ۲ evaporation at end of dry season
- Still just small improvements

Methods

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Improvements by the hydrology:

Motivation

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Dynamic groundwater tables do not show a strong improvement

See more ... Howard Springs AdelaideRiver DalyUncleared DryRiver **SturtPlains**

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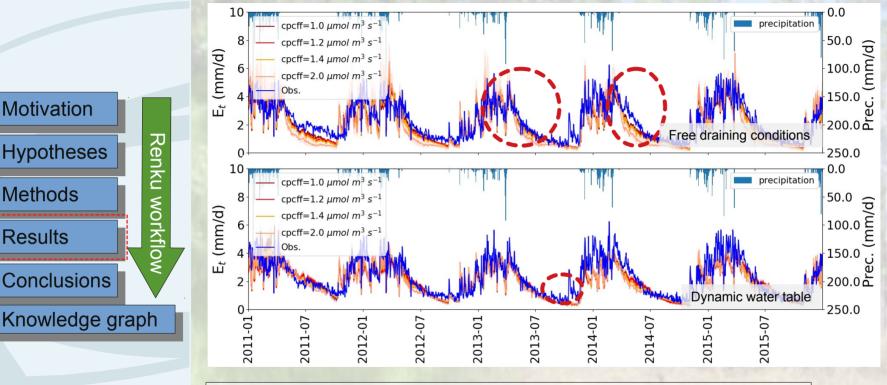
Motivation

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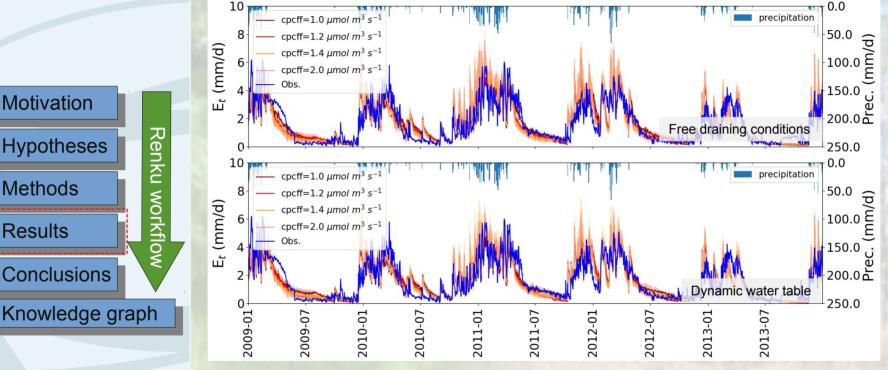


- Higher values of the water transport cost parameter improve assimilation
 - Dynamic groundwater tables do not help for assimilation

See more ... Howard Springs AdelaideRiver DalyUncleared **DryRiver SturtPlains**

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Improvements by the hydrology:

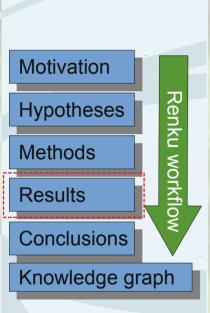
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 Dynamic groundwater tables do not show strong improvements

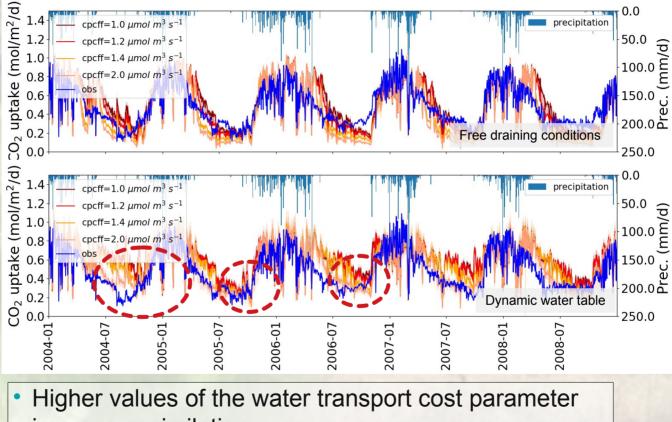
HYDROLOGY AND CARBON COSTS





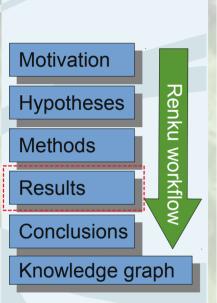
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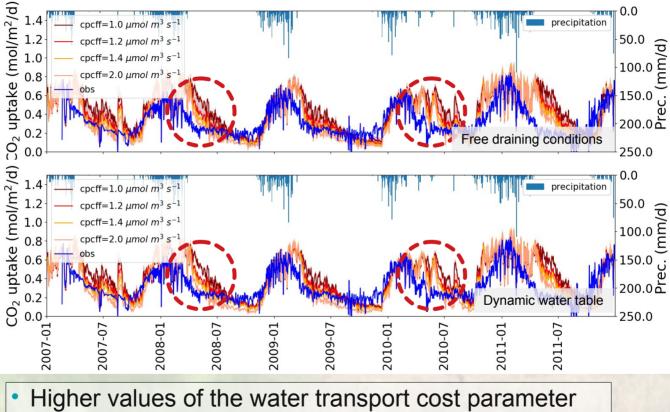
- improve assimilation
- Dynamic groundwater tables do not help for assimilation





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- Higher values of the water transport cost parameter improve assimilation
- Dynamic groundwater tables do not help for assimilation



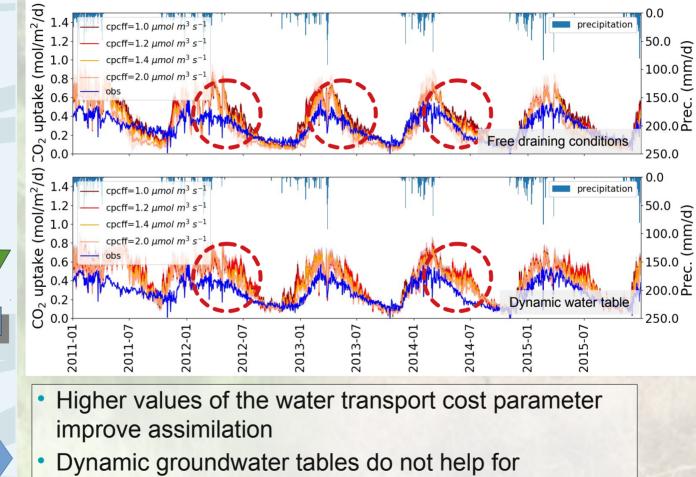
Motivation Hypotheses Methods Results Conclusions Knowledge graph

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assimilation





 $(mol/m^2/d)$ T 0.0 cpcff=1.0 μ mol m³ s⁻¹ precipitation ^{50.0} (p/uu) cpcff=1.2 μ mol m³ s⁻¹ cpcff=1.4 μ mol m³ s⁻¹ 1.0cpcff=2.0 μ mol m³ s⁻¹ 0.8 uptake 0.6 150.0 **Motivation** بو 200.0 م 0.4 0.2 C02 Free draining conditions Renku workflow **Hypotheses** 250.0 0.0 $(mol/m^2/d)$ 0.0 cpcff=1.0 μ mol m³ s⁻¹ precipitation **Methods** 50.0 (p/uu) 100.0 u) 150.0 u 1.2 cpcff=1.2 μ mol m³ s⁻¹ cpcff=1.4 μ mol m³ s⁻¹ 1.0 Results cpcff=2.0 μ mol m³ s⁻¹ 0.8 بة 200.0 J Conclusions Dynamic water table Knowledge graph 250.0 -01 2009-01 2010-01 -07 2011-01 2011-07 -07 -01 -07 2009-07 Ó 2 m 3 201 201 201 201 201 Higher values of the water transport cost parameter

- improve assimilation
 - Dynamic groundwater tables do not help for assimilation

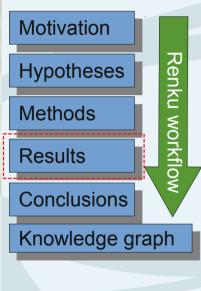
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PERFORMANCES

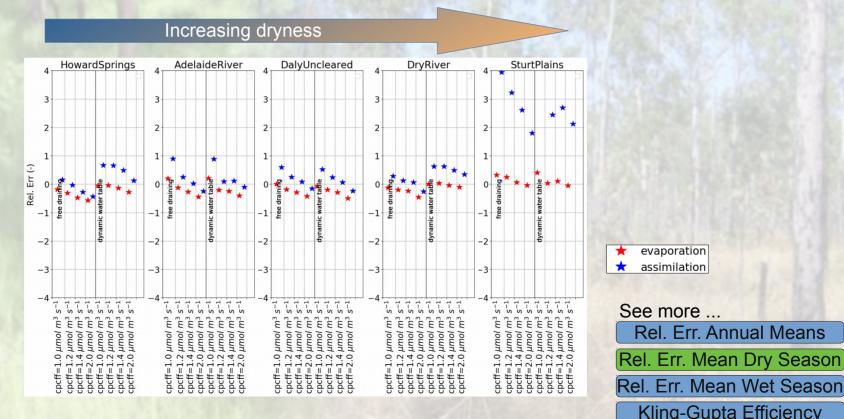
- Assimilation increasingly over-estimated for drier areas
- Evaporation still okay



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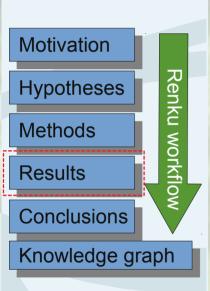


evaporation assimilation

Kling-Gupta Efficiency

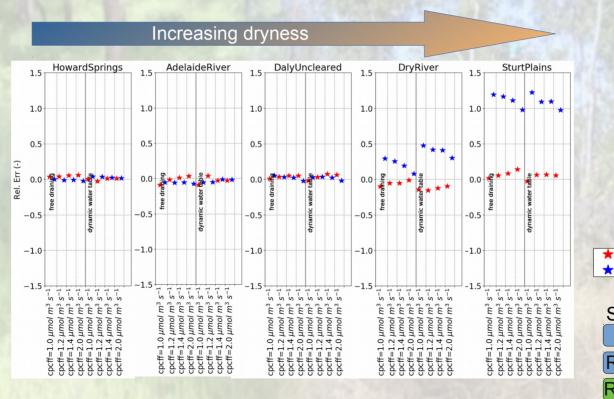
PERFORMANCES

- Assimilation increasingly over-estimated for drier areas
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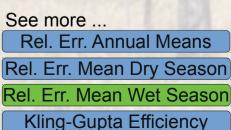


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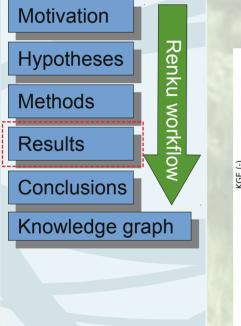


evaporation

assimilation

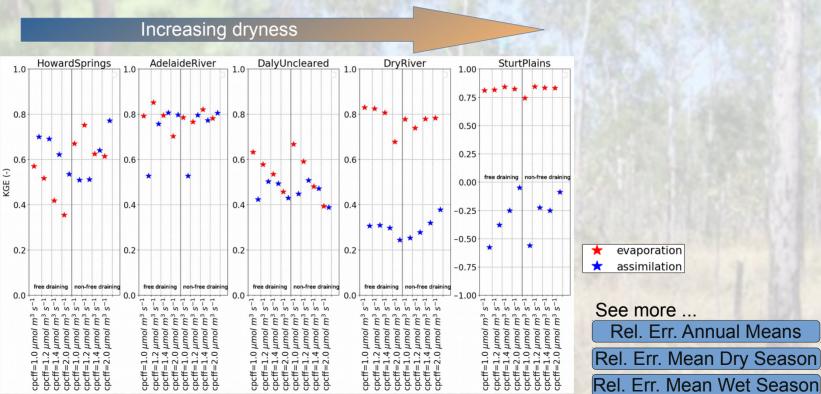
PERFORMANCES

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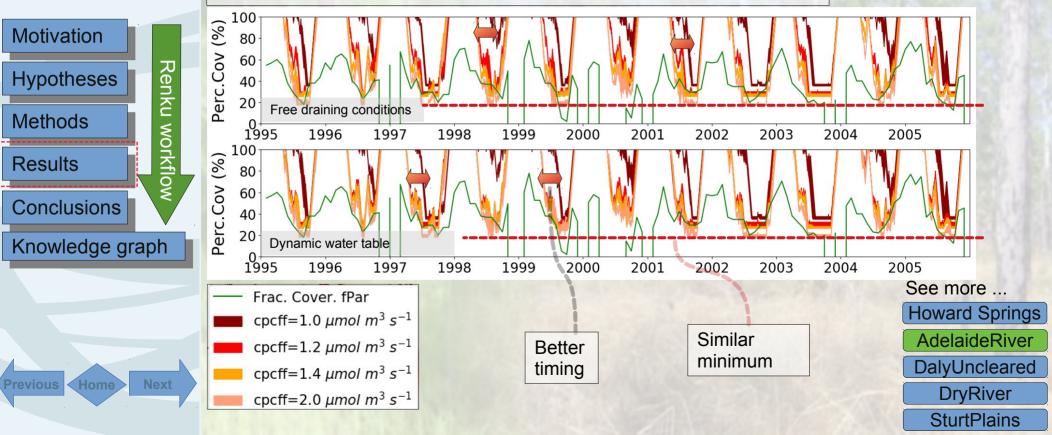
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Kling-Gupta Efficiency



VEGETATION DYNAMICS

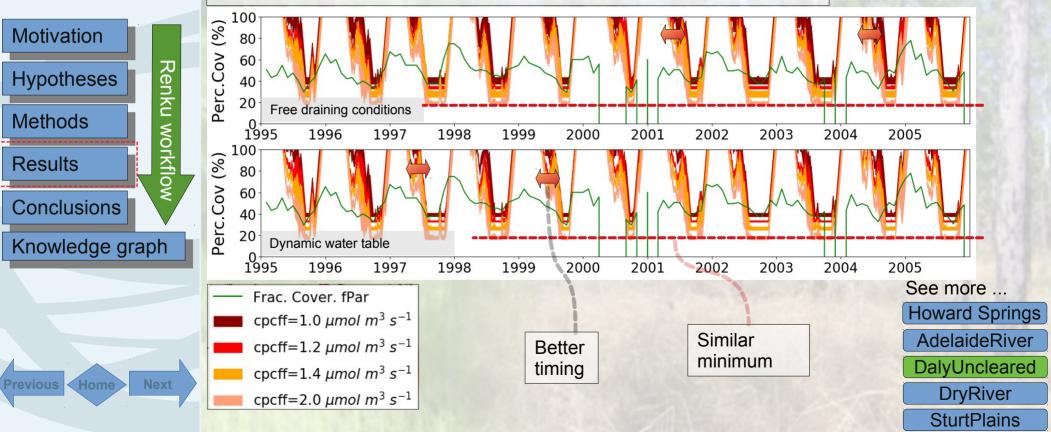
- Temporal signal largely reproduced
- Timing improves for higher cost factor
- Similar minimum cover for dynamic water tables

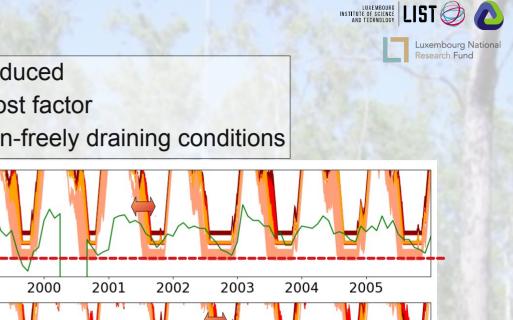




VEGETATION DYNAMICS

- Temporal signal largely reproduced
- Timing improves for higher cost factor
- Similar minimum cover for dynamic water tables

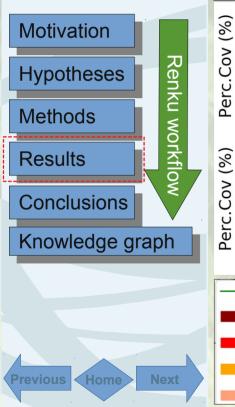


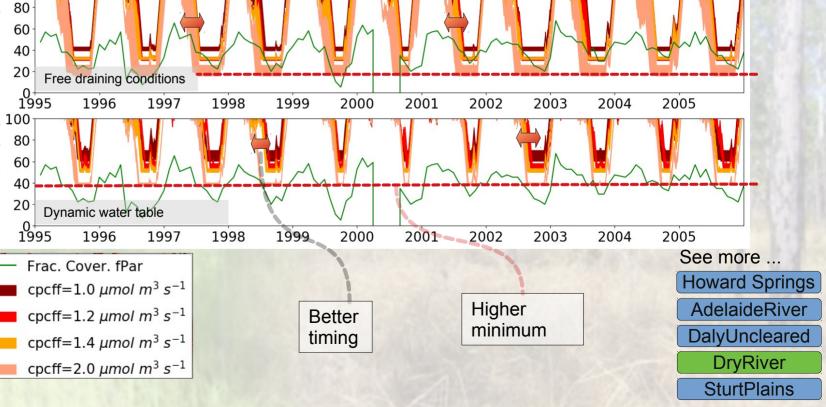


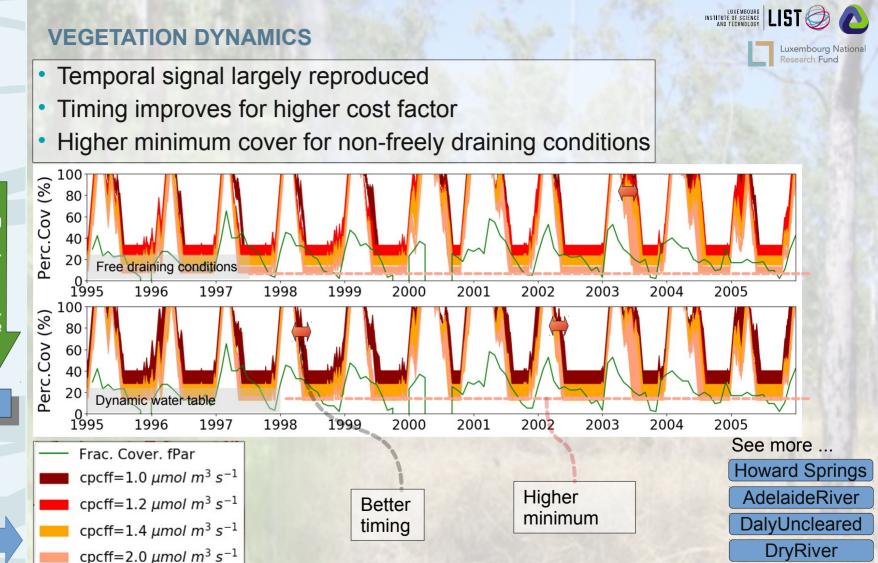
VEGETATION DYNAMICS

100

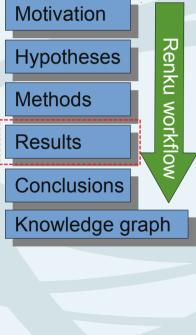
- Temporal signal largely reproduced
- Timing improves for higher cost factor
- Higher minimum cover for non-freely draining conditions







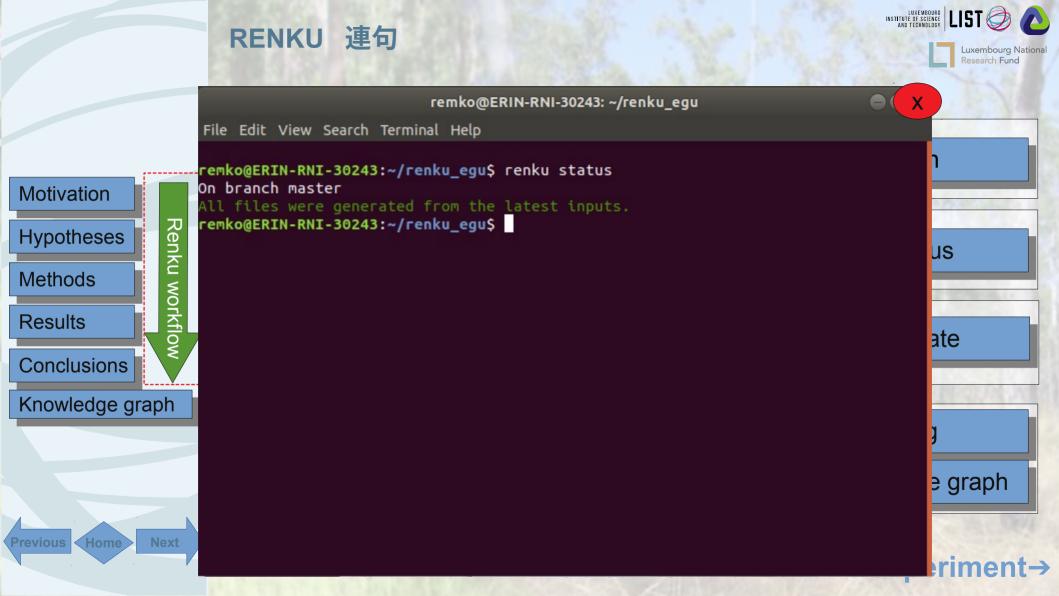
SturtPlains

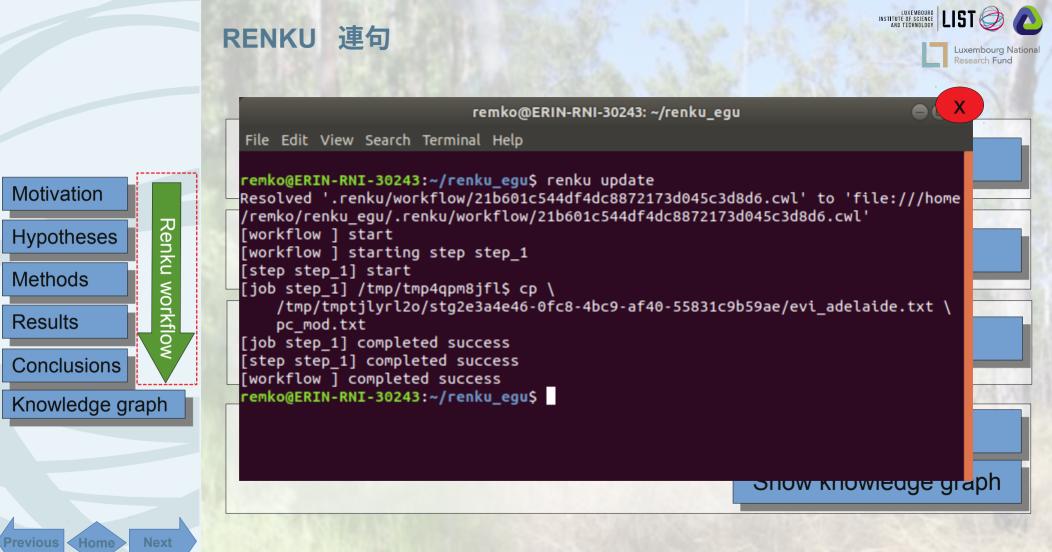


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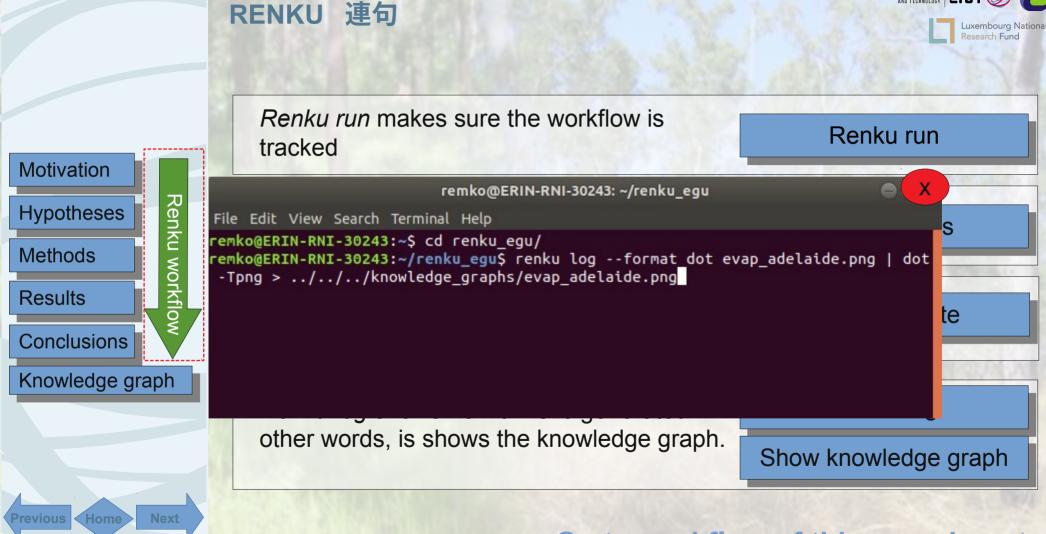
Next







Go to workflow of this experiment→



Go to workflow of this experiment→

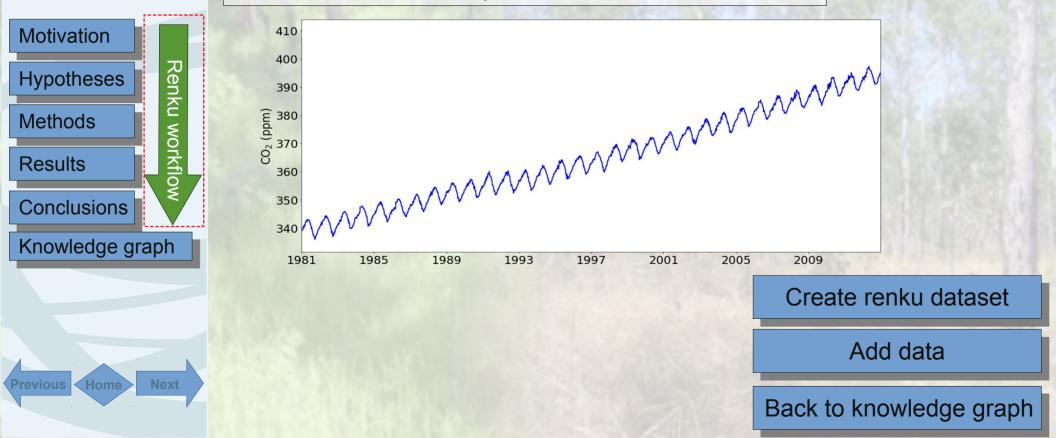
INSTITUTE OF SCIENCE AND TECHNOLOGY

CO₂ RECORDS

Atmospheric CO₂ levels are needed as input for the VOM model. Therefore, weekly data is taken from the Mauna Loa observatory.

INSTITUTE OF SCIENCE AND TECHNOLOGY

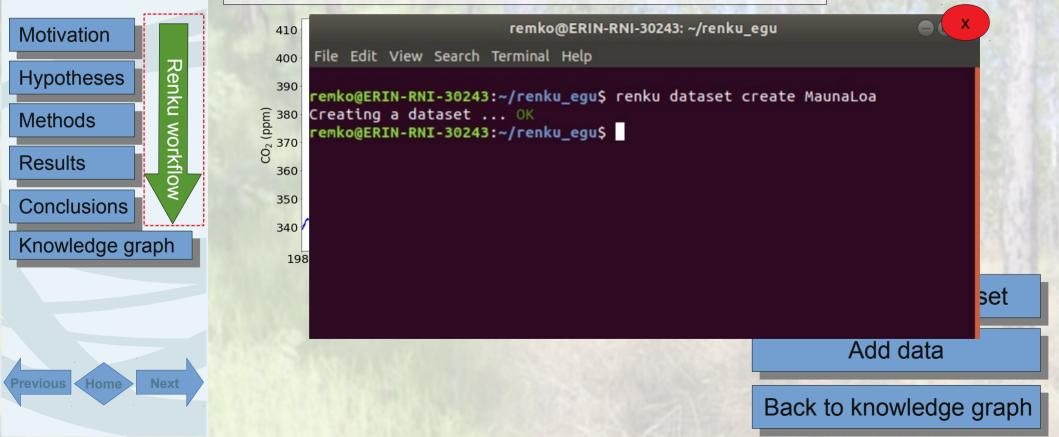
> Luxembourg National Research Fund



CO₂ RECORDS

INSTITUTE OF SCIENCE AND FERMILLER LLIST OF CONTRACTOR LUXEMBOURG LLIST OF CONTRACTOR LUXEMBOURG LLIST OF CONTRACTOR LUXEMBOURG

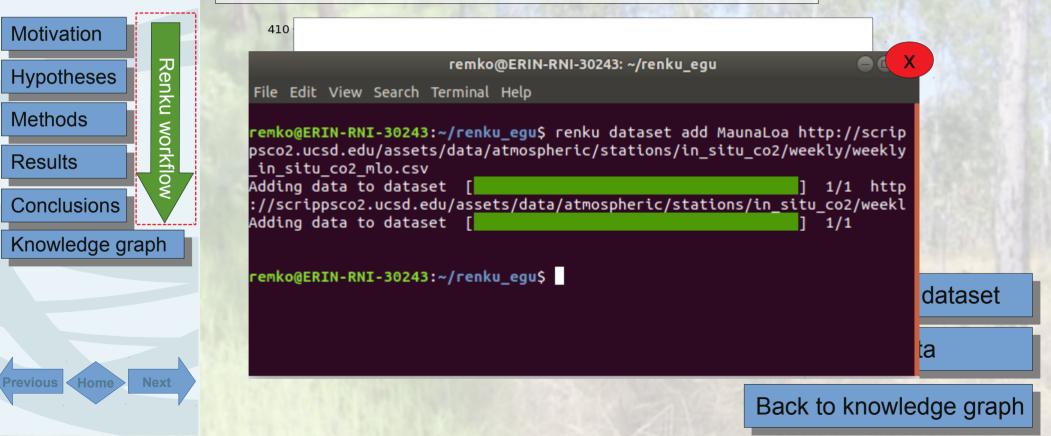
Atmospheric CO₂ levels are needed as input for the VOM model. Therefore, weekly data is taken from the Mauna Loa observatory.



CO₂ RECORDS

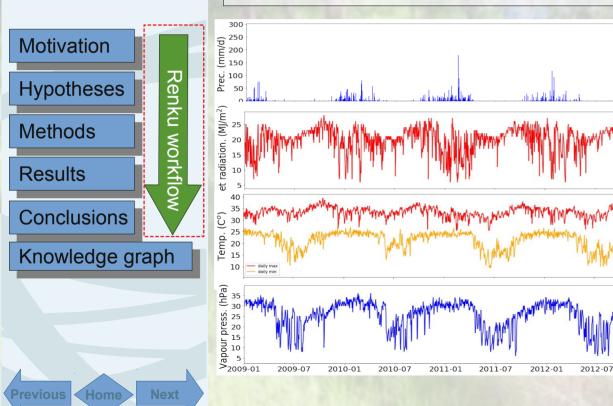


Atmospheric CO₂ levels are needed as input for the VOM model. Therefore, weekly data is taken from the Mauna Loa observatory.



SILO WEATHERDATA

Meteorological data are needed as input for the VOM model. Data is taken from the Australian Silo weatheroffice.



Create renku dataset

Add data

Back to knowledge graph



SILO WEATHERDATA

Temp. (C^o)

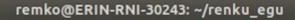
(hPa)

Vapour

Renku workflow

Next

Meteorological data are needed as input for the VOM model. Data is taken from the Australian Silo weatheroffice.



File Edit View Search Terminal Help

```
remko@ERIN-RNI-30243:~/renku_egu$ renku dataset create SILO
Creating a dataset ... OK
remko@ERIN-RNI-30243:~/renku_egu$
```



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u dataset

Knowledge graph

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Conclusions

Motivation

Hypotheses

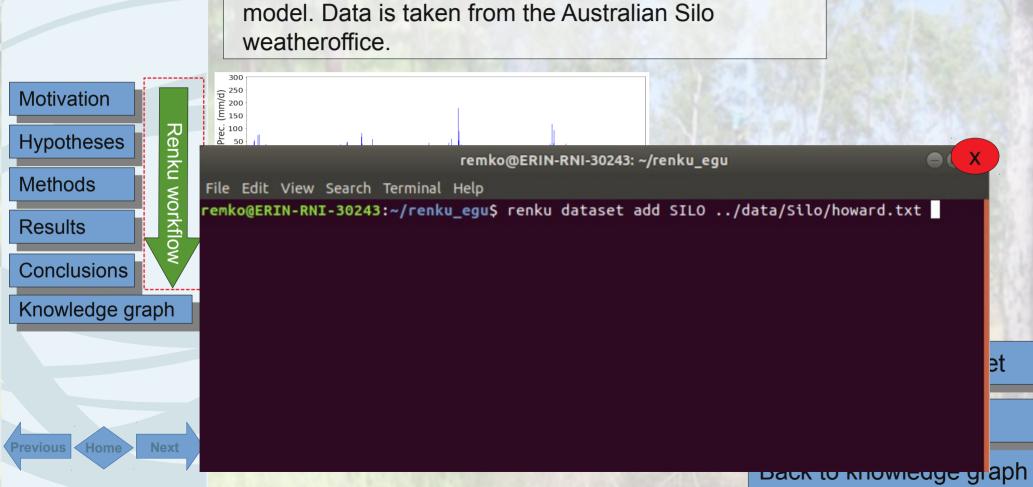
Methods

Results

SILO WEATHERDATA

Meteorological data are needed as input for the VOM





CREATE MODEL INPUT



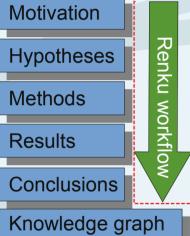
The meteorological data and the atmospheric CO_2 data need to be processed and formatted in order to serve as input for the model.

	0.000	Æ				dailyw	veather.pr	n		Save	
	Open 🕶									Save	
	Dcum	Day	Month	Үеаг	T.Max	T.Min	Rain	Radn	VP Pres	Ca	
	1	1	1	1980	31.50	23.50	10.70	10.00	30.00 1005.00	337.67	
	2	2	1	1980	28.50	23.50	40.90	6.00	29.00 1004.00	337.67	
	3	3	1	1980	28.50	23.50	14.10	7.00	30.00 1004.50	337.67	
	4	4	1	1980	30.00	23.00	14.30	10.00	31.00 1004.50	337.67	
	5	5	1	1980	31.00	24.00	17.60	12.00	31.00 1005.00	337.67	
	6	6	1	1980	31.50	24.50	12.90	14.00	33.00 1004.00	337.48	
s	7	7	1	1980	32.00	24.00	85.20	16.00	31.00 1005.50	337.48	
	8	8	1	1980	33.50	24.00	30.50	24.00	34.00 1006.50	337.48	
	9	9	1	1980	32.50	22.50	17.30	17.00	31.00 1006.50	337.48	
	10	10	1	1980	32.50	24.50	8.10	20.00	31.00 1007.00	337.48	
	11	11	1	1980	34.00	25.00	16.20	22.00	33.00 1008.50	337.48	
	12	12	1	1980	32.50	26.50	11.20	14.00	34.00 1009.00	337.48	
	13	13	1	1980	34.00	25.50	17.90	19.00	33.00 1007.50	338.38	
- S	14	14	1	1980	34.50	24.00	26.80	15.00	32.00 1008.50	338.38	
	15	15	1	1980	33.50	24.00	26.10	20.00	34.00 1009.00	338.38	
rkf	16	16	1	1980	34.00	26.00	2.50	21.00	34.00 1008.50	338.38	
	17	17	1	1980	32.50	21.50	20.10	14.00	29.00 1009.00	338.38	
	18	18	1	1980	29.00	23.50	25.70	6.00	30.00 1009.00	338.38	
	19	19	1	1980	31.50	24.50	8.10	16.00	31.00 1008.00	338.38	
	20	20	1	1980	33.00	23.50	9.50	20.00	32.00 1009.50	338.49	
$ \langle \langle \rangle \rangle > \langle \rangle$	21	21	1	1980	32.50	24.00	41.60	17.00	30.00 1010.00	338.49	
S	22	22	1	1980	31.50	22.00	10.70	17.00	26.00 1009.00	338.49	
3	23	23	1	1980	32.00	23.50	9.10	18.00	29.00 1009.00	338.49	
	24	24	1	1980	31.50	24.50	9.00	16.00	32.00 1008.50	338.49	
	25	25	1	1980	32.00	24.50	1.80	19.00	32.00 1008.50	338.49	
	26	26	1	1980	31.00	24.50	6.50	16.00	31.00 1007.50	338.49	
graph	27	27	1	1980	31.00	23.50	20.00	10.00	31.00 1005.50	338.05	
graph	28	28	1	1980	29.50	23.50	23.40	11.00	30.00 1005.00	338.05	
	29	29	1	1980	31.00	24.50	6.20	14.00	32.00 1005.50	338.05	
	30	30	1	1980	30.00	25.50	9.70	10.00	30.00 1006.50	338.05	
	31	31	1	1980	32.50	24.50	3.80	10.00	33.00 1008.50	338.05	
	32	1	2	1980	31.00	24.50	20.10	12.00	33.00 1010.00	338.05	
	33	2	2	1980	30.50	22.50	38.40	9.00	30.00 1009.50	338.05	
	34	3	2	1980	30.50	23.50	14.00	12.00	30.00 1006.50	338.14	
	25	л	2	1000	27 50	24 66	11 00	7 66	30 00 100/ E0	220 14	

Run command

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Back to knowledge graph



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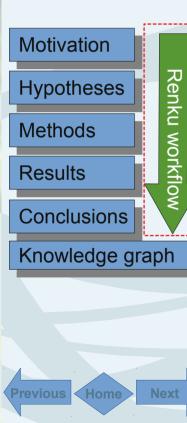
Next

CREATE MODEL INPUT

The meteorological data and the atmospheric CO

remko@ERIN-RNI-30243: ~/renku_egu

File Edit View Search Terminal Help



-emko@ERIN-RNI-30243:~/renku_egu\$ renku run python3 src_py/write_dailyweather.py
 -im data/Silo/howard.txt -ic data/MaunaLoa/weekly_in_situ_co2_mlo.csv -p linear
 -o data/VOM_input/HowardSprings/dailyweather.prn

hand

dge graph

X

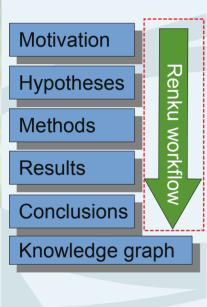
Luxembourg National Research Fund

RUN VOM OPTIMIZATION



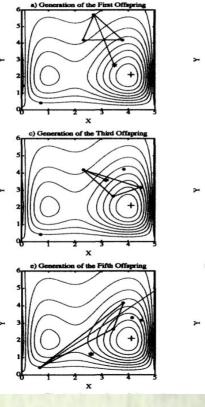
The Shuffled Complex Evolution algorithm is used in order to derive the vegetation properties that maximize NCP.

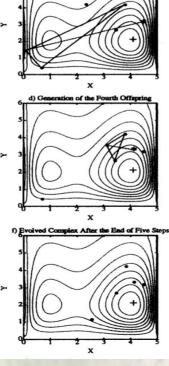
b) Generation of the Second Offsprin



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Next





- Sample s points
- Rank points
- Partition into complexes
- Evolve complex
- Shuffle complexes
- Check convergence

Run command

Back to knowledge graph

Figure from : Duan, Q., Sorooshian, S., Gupta, V.K., 1994. Optimal use of the SCE-UA global optimization method for calibrating watershed models. Journal of Hydrology 158, 265–284. https://doi.org/10.1016/0022-1694(94)90057-4

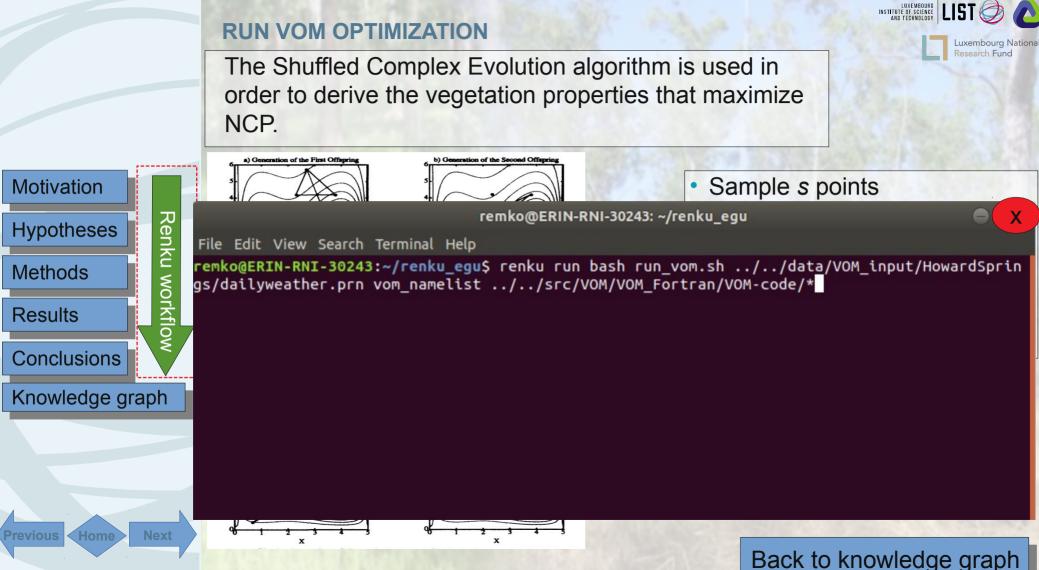
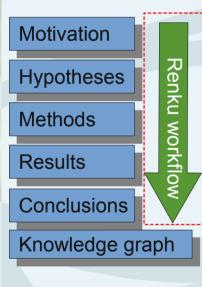


Figure from : Duan, Q., Sorooshian, S., Gupta, V.K., 1994. Optimal use of the SCE-UA global optimization method for calibrating watershed models. Journal of Hydrology 158, 265-284. https://doi.org/10.1016/0022-1694(94)90057-4

UNCERTAINTY ESTIMATES & STATISTICS



The model runs with the 5% highest NCP are selected in order to construct uncertainty bounds.



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Next

To assess the model performance, several independent datasets are used:

- DINGO fluxdata
- Fraction vegetation cover from fPar

The performances are assessed by:

- Timeseries with uncertainties
- Relative errors annuan and seasonal means
- Kling-Gupta efficiencies
- Residuals

Run command

Back to knowledge graph

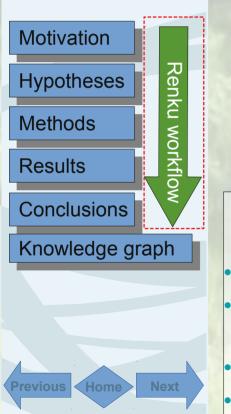
UNCERTAINTY ESTIMATES & STATISTICS



X

The model runs with the 5% highest NCP are selected in order to construct uncertainty bounds.

remko@ERIN-RNI-30243: ~/renku_egu



R

m

Residuals

File Edit View Search Terminal Help

Kling-Gupta efficiencies

remko@ERIN-RNI-30243:~/renku_egu\$ renku run python3 src_py/sce_uncertainty.py -i
 data/VOM_output/HowardSprings/sce_out.txt -o data/VOM_output/HowardSprings/free
 drainage_cpcff1.0/bestruns/ -p -w work/HowardSprings/freedrainage_cpcff1.0/best/
 5 -op 1 1 1 1 1 0 1 -d data/VOM_input/HowardSprings/dailyweather.prn -eo data
/DINGO/Ea_howard.txt -ea data/DINGO/GPPdaily_howard.txt

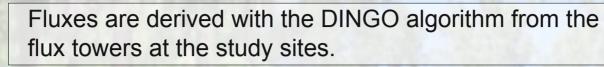
Run command

Back to knowledge graph

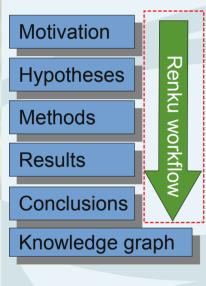
DINGO

10



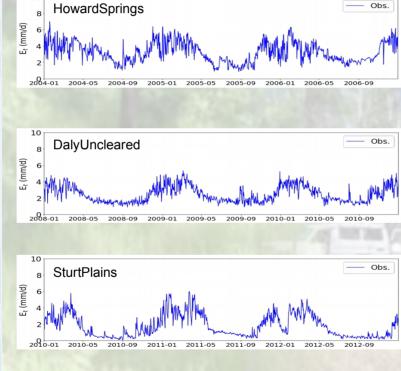


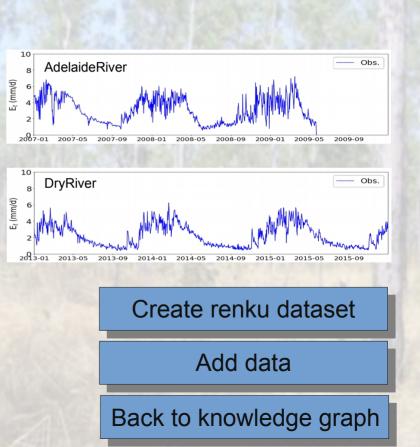
Obs.

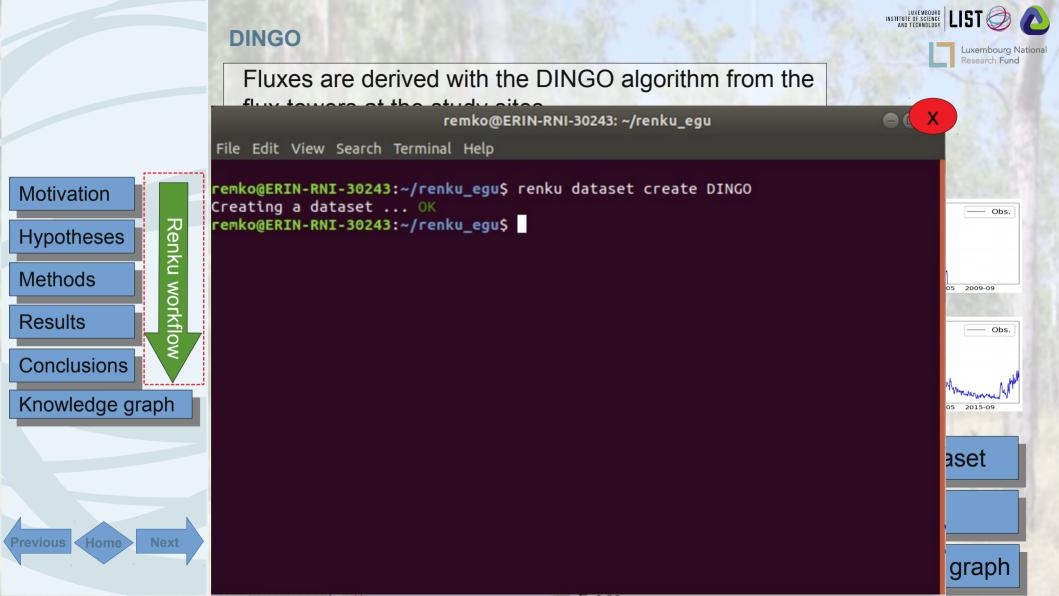


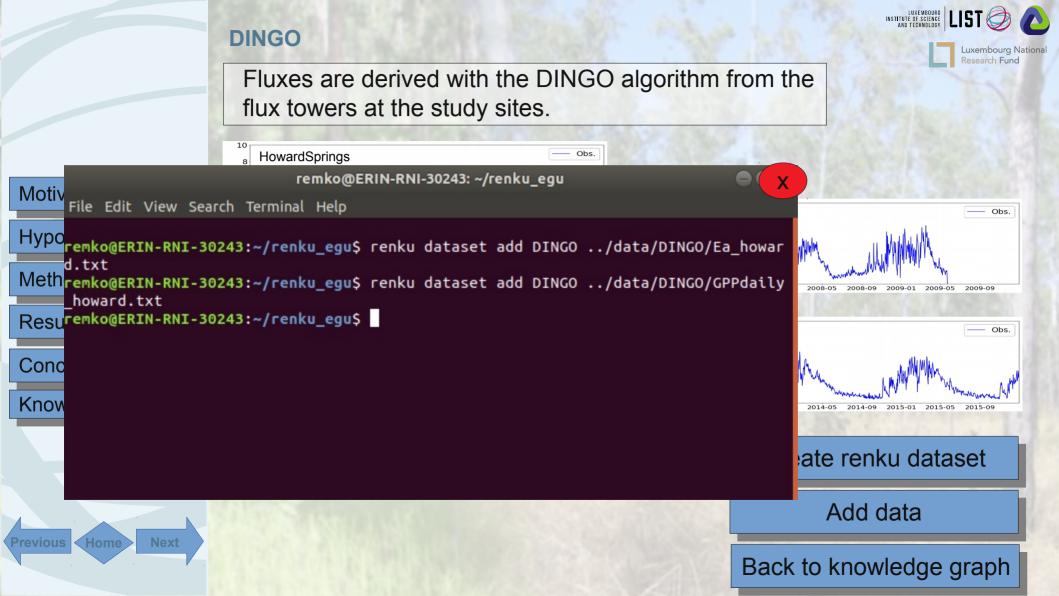
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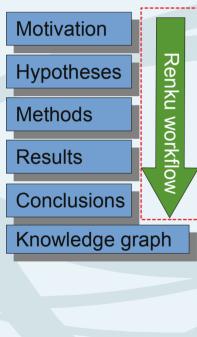




FRACTIONAL COVER

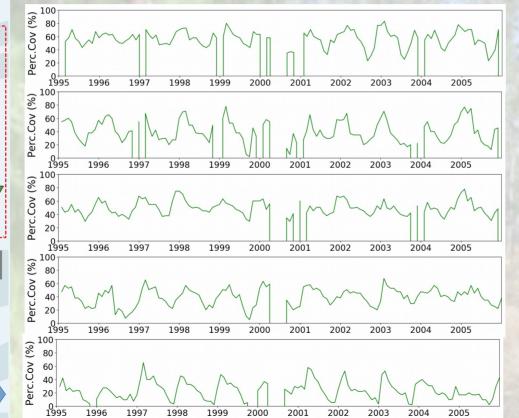
The Enhanced Vegetation Index from MODIS is used to compare the modelled and observed vegetation dynamics.





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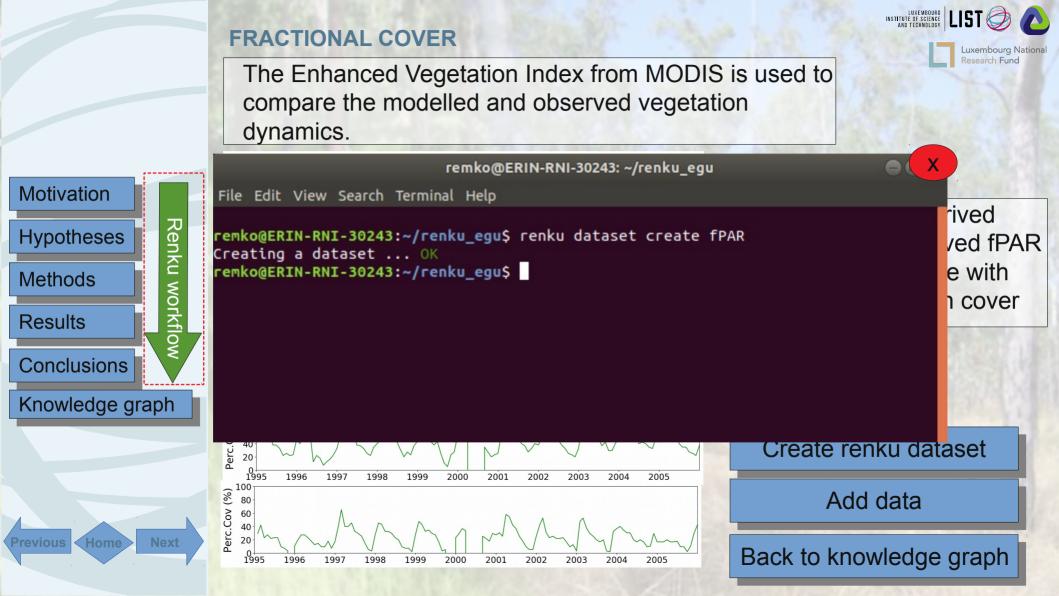


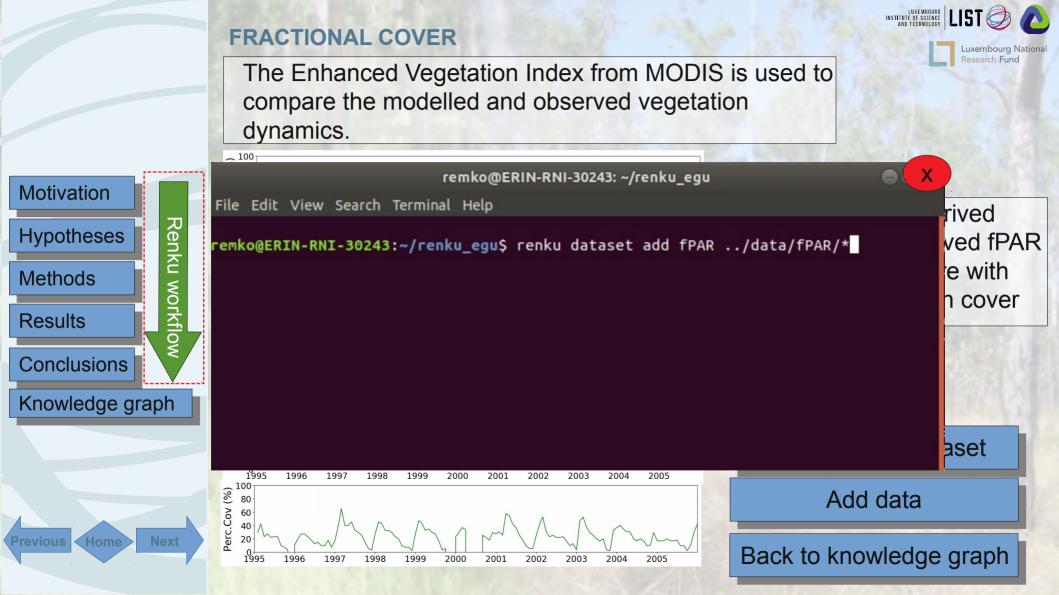
Fractional cover derived from satellite observed fPAR are used to compare with modelled vegetation cover

Create renku dataset

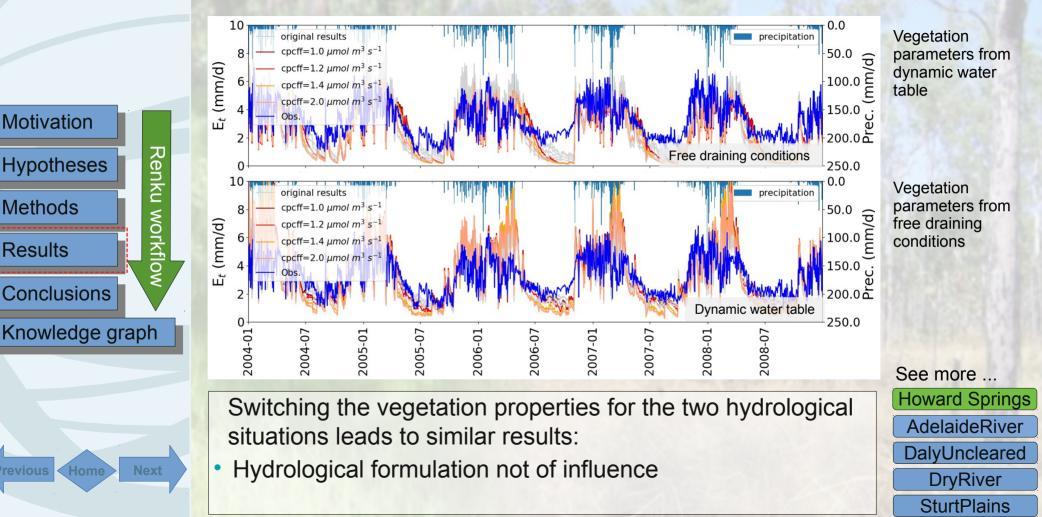
Add data

Back to knowledge graph



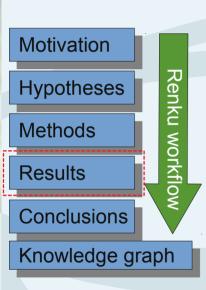






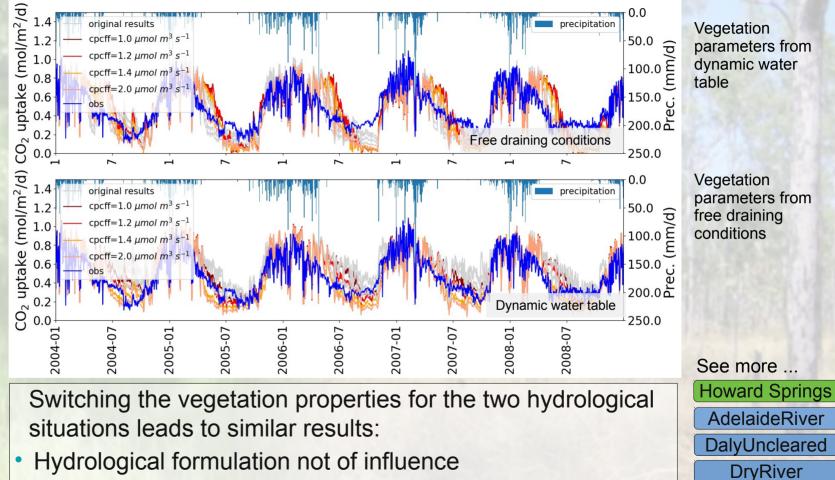


SturtPlains



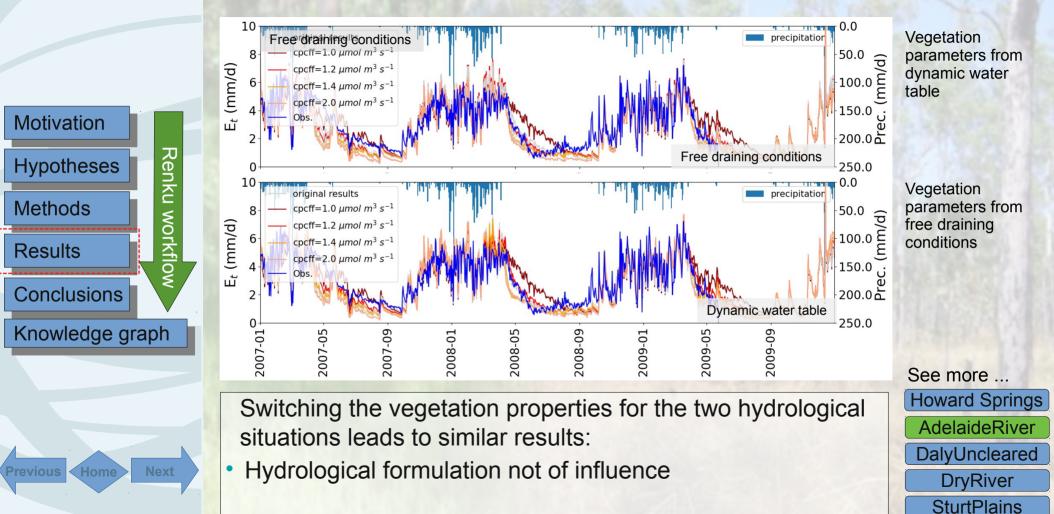
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Hydrological formulation not of influence



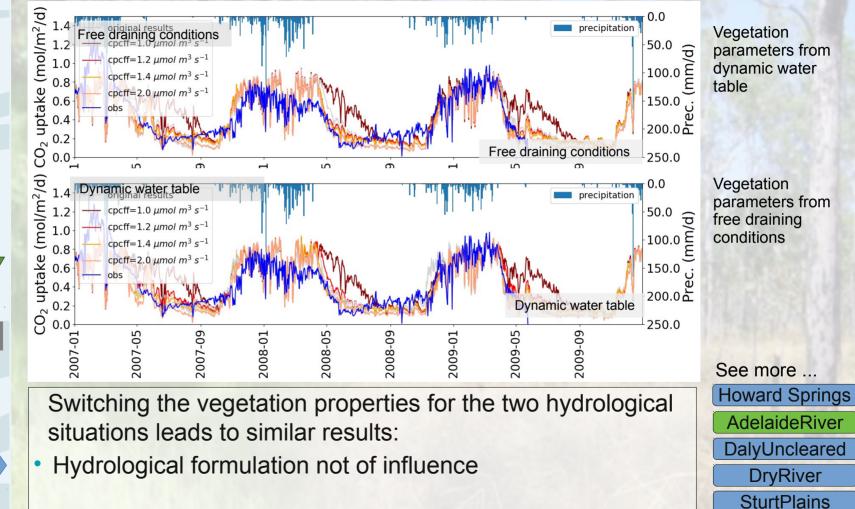




Motivation Hypotheses Methods Results Conclusions Knowledge graph

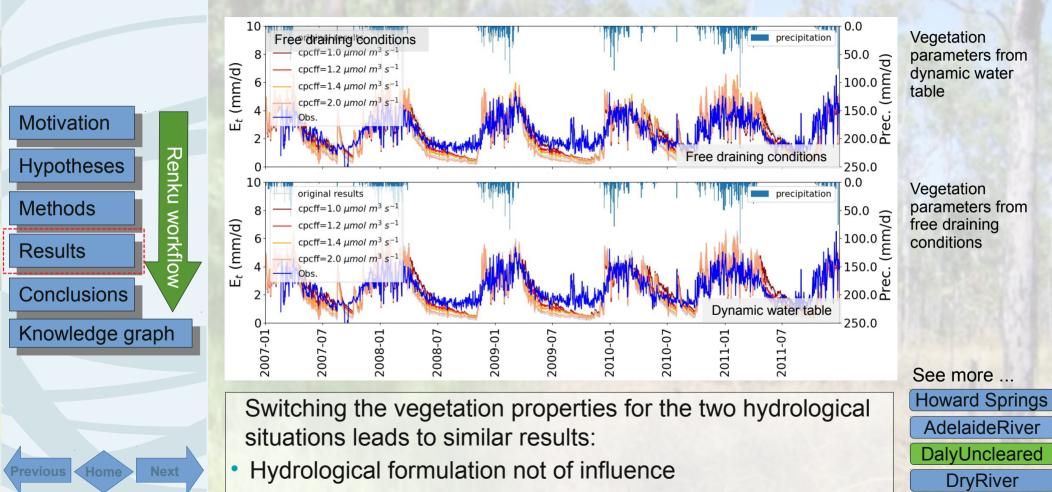
Previous < Home

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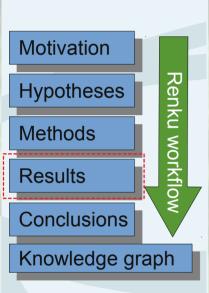


SturtPlains



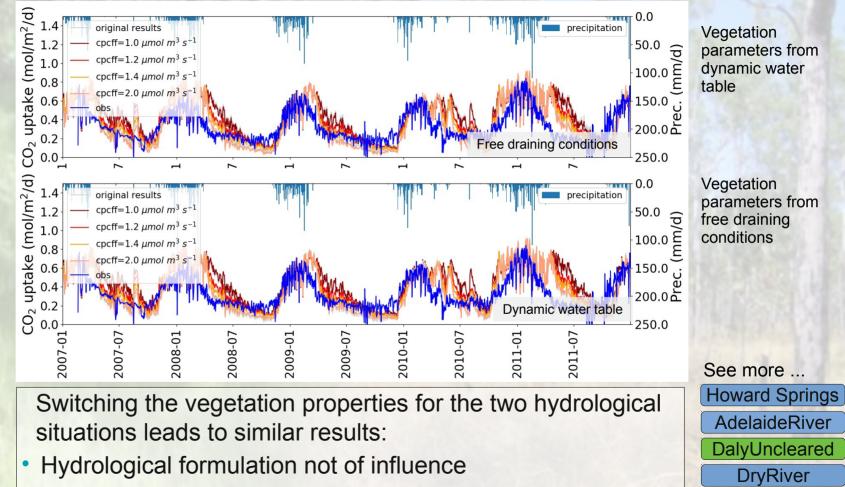


SturtPlains



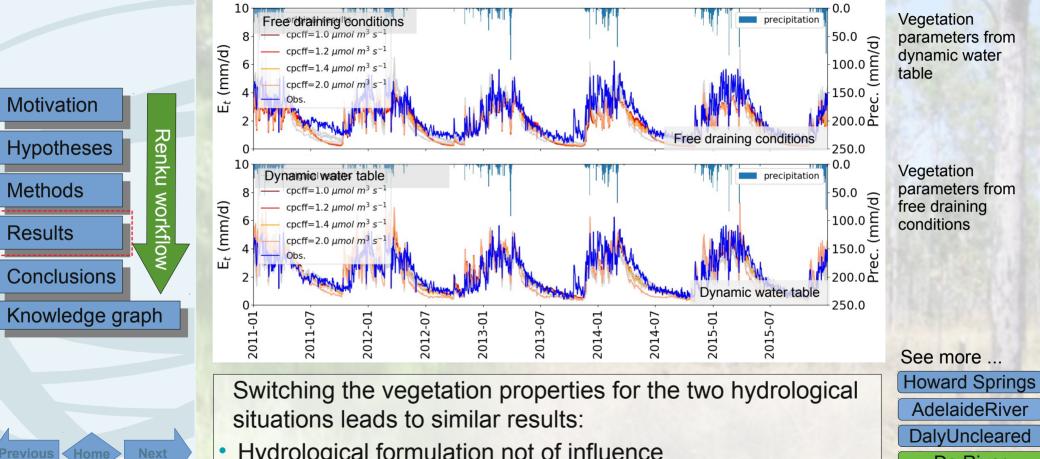
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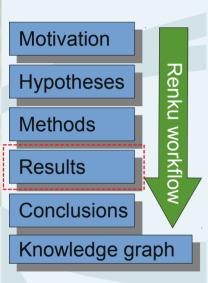
DryRiver SturtPlains



Hydrological formulation not of influence

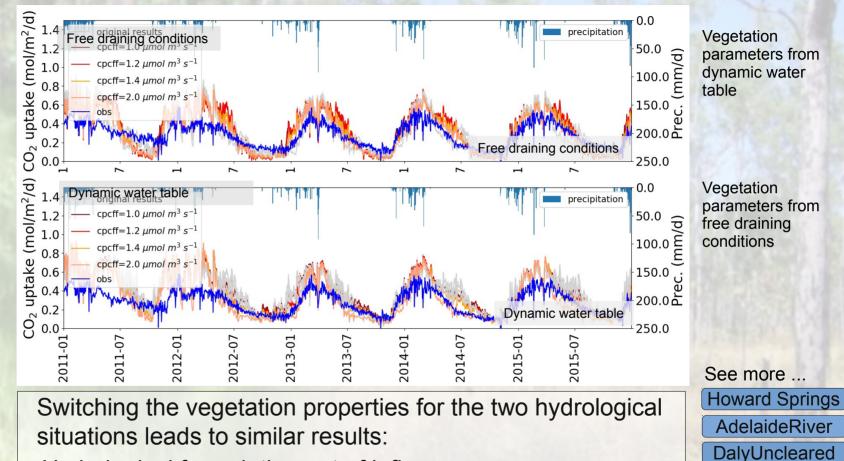


DryRiver SturtPlains



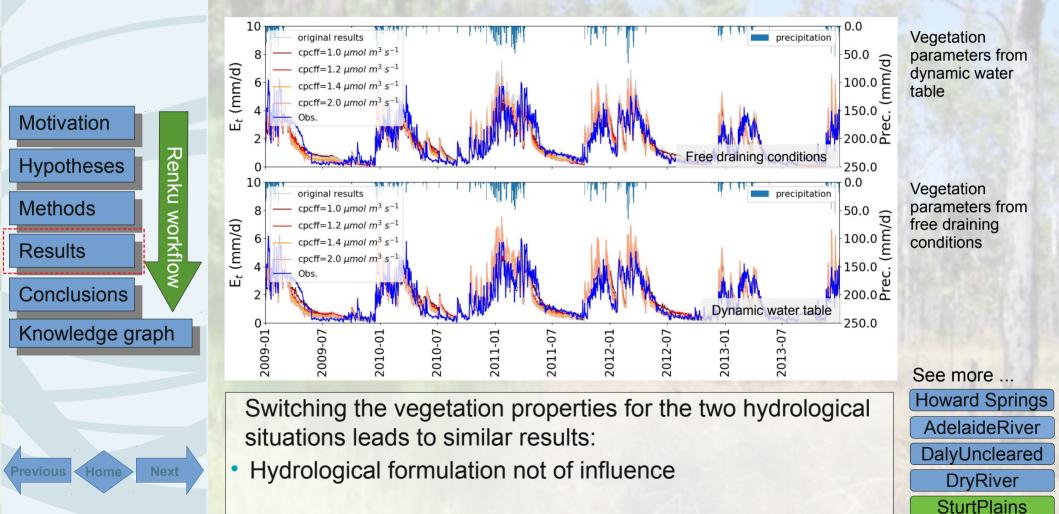
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Hydrological formulation not of influence

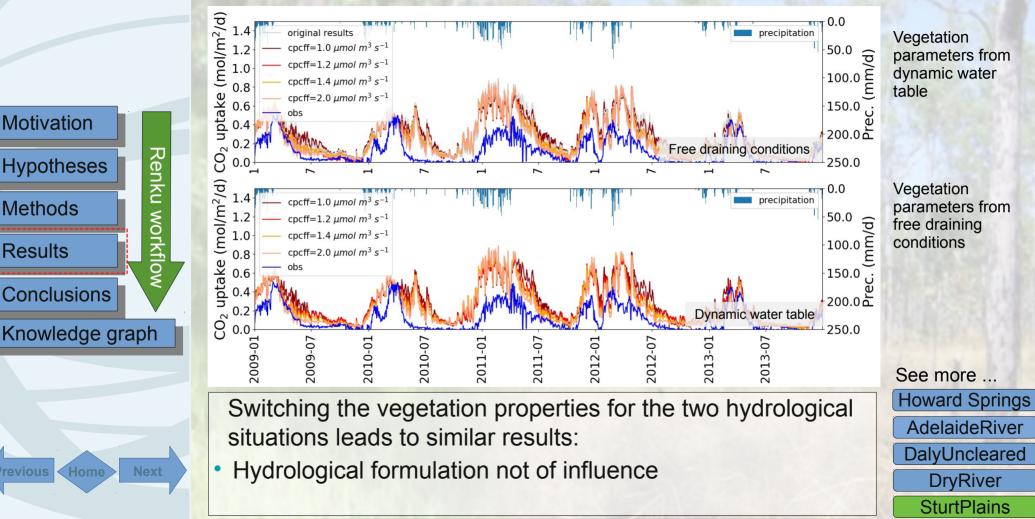




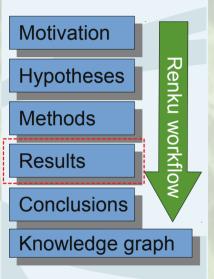
Methods

Results



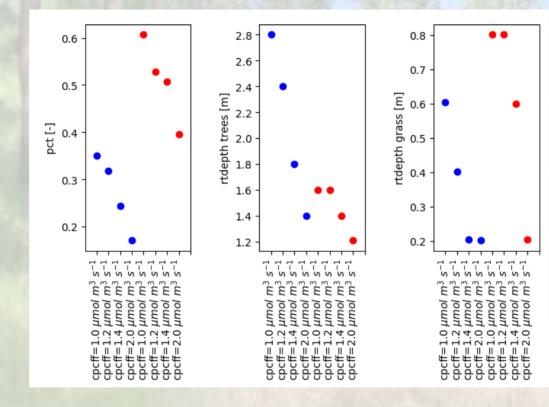


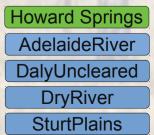




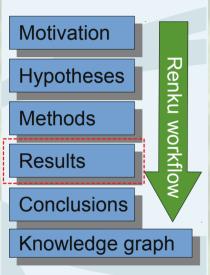
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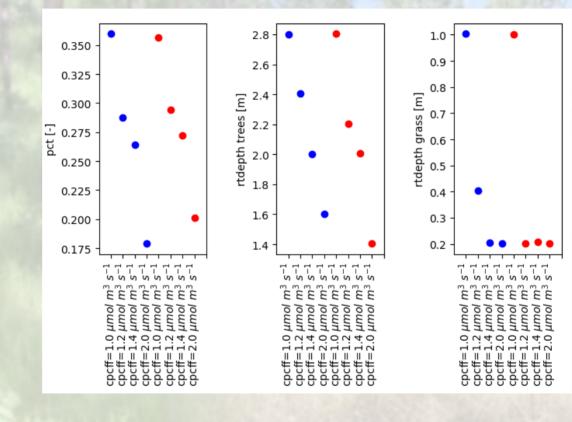
Next

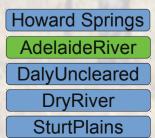






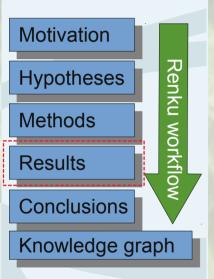






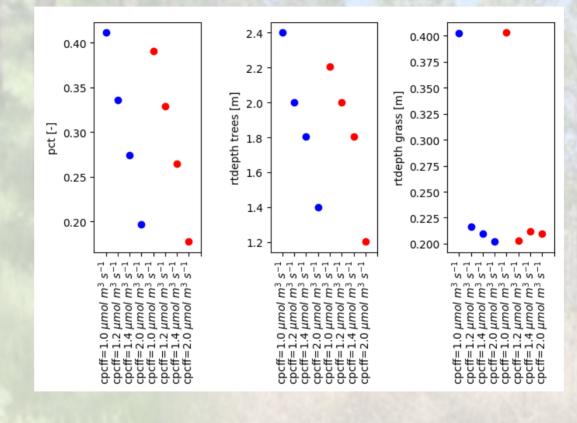






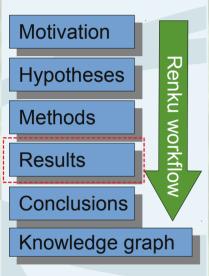
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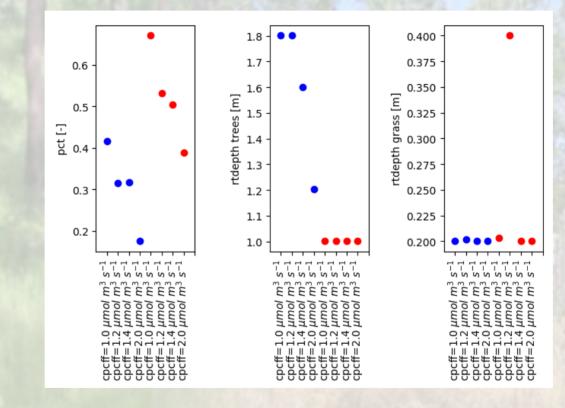
Next



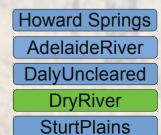
Howard Springs AdelaideRiver DalyUncleared DryRiver SturtPlains



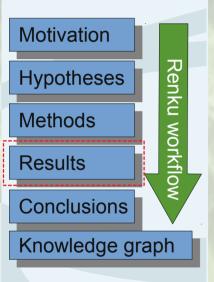


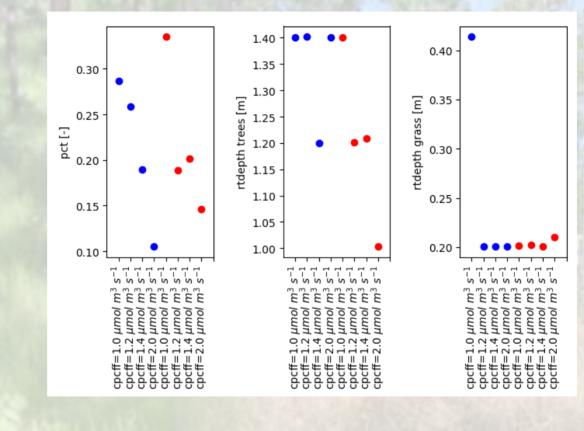












Howard Springs AdelaideRiver DalyUncleared DryRiver SturtPlains

