



Universidad
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ACMA
Agua, Clima y Medio Ambiente



HML - UAH
Hydrological Modelling Lab

SUNTORY



A new calibration strategy within SWATplusR: The Guajaraz River basin study case

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Dpt. of Geology, Geography and Environment

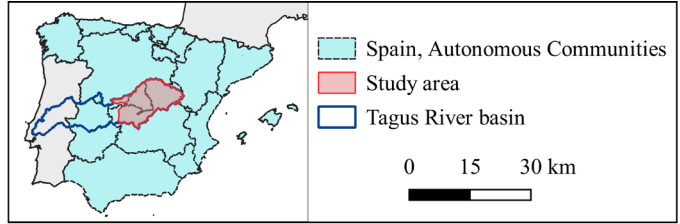
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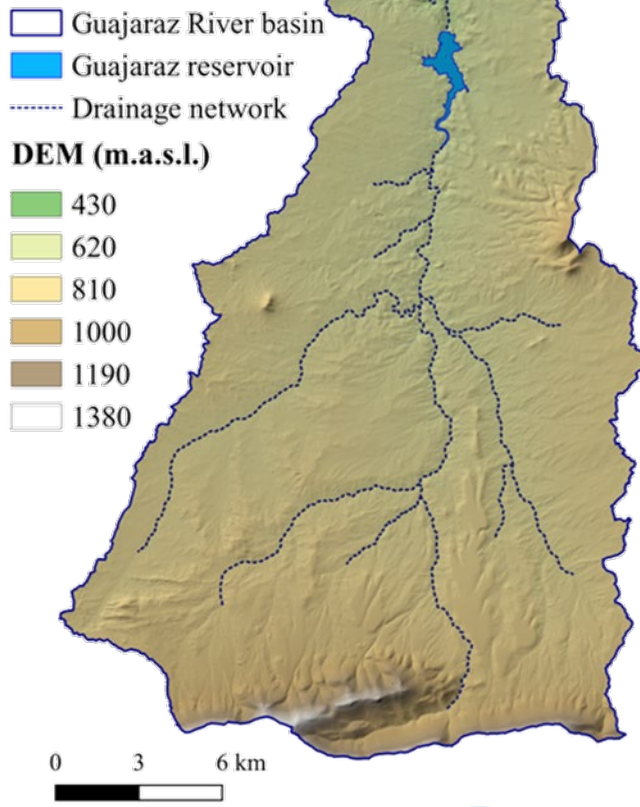
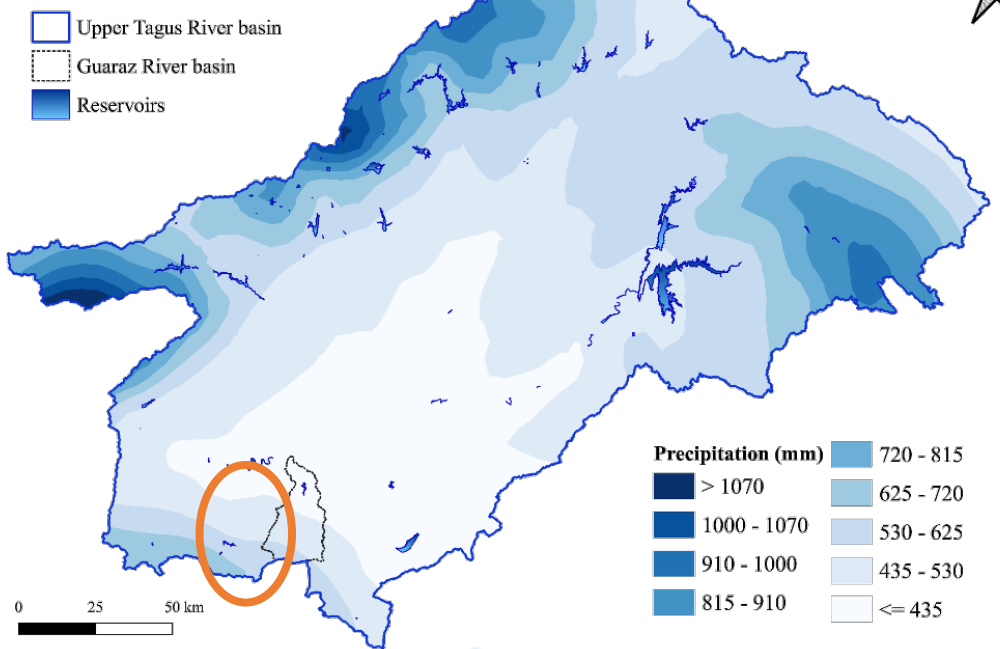
INTRODUCTION



Working context



Average annual precipitation (mm, 1951-2019) in the upper sector of the Tagus River basin and in the Guajaraz River basin






















Area: 417 km²

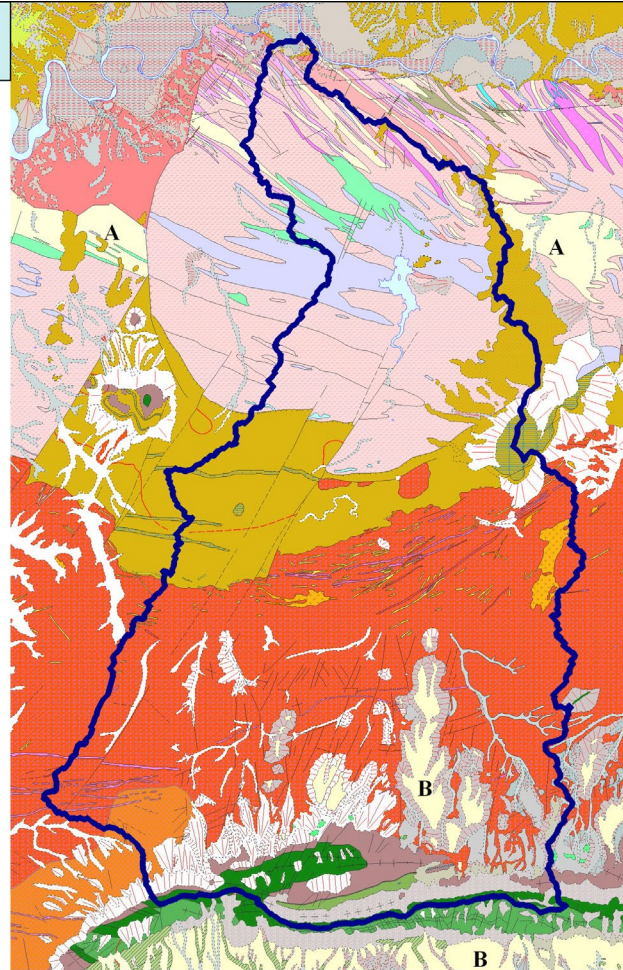
Average precipitation:

1951 - 2019: **465 mm**
1980 - 2019: **441 mm**

Suntory's concern about water resources availability

Geology of the Guajaraz River basin

- | | |
|--|--|
|  Biotitic granitoids. Equigranular facies |  Pebbles, gravels and boulders with sandy clay matrix |
|  Biotitic granitoids. Porphyritic facies. Medium-coarse-grained |  Pebbles, gravels, sands and clays |
|  Fm PUSA: Marbles |  Purple Series: Sandstones, quartzites and shales |
|  Fm PUSA: Slates, banded slates and siltstones, carbonate levels |  Schists, porphyritic metarenitic, paragneiss and calcium silicate rocks. |
|  Green shales, greywackes, siltstones, sandstones, and limestones |  B Siliceous orthoconglomerate with sandy-clay matrix |
|  Hillside landslides. Angular cobbles, sandy-clayey matrix. |  Silt, grey and sand |
|  Inhomogeneous granitoids. Biotitic facies |  Thick-banked quartzite: ARMORICANA |
|  Intermediate and basic rocks. Granodiorites and tonalites |  Two-mica granitoids. Porphyritic facies. Medium-fine-grained |
|  A Leucogneisses |  Water bodies |
|  Mesocrate-melanocrate glandular orthogneisses | |

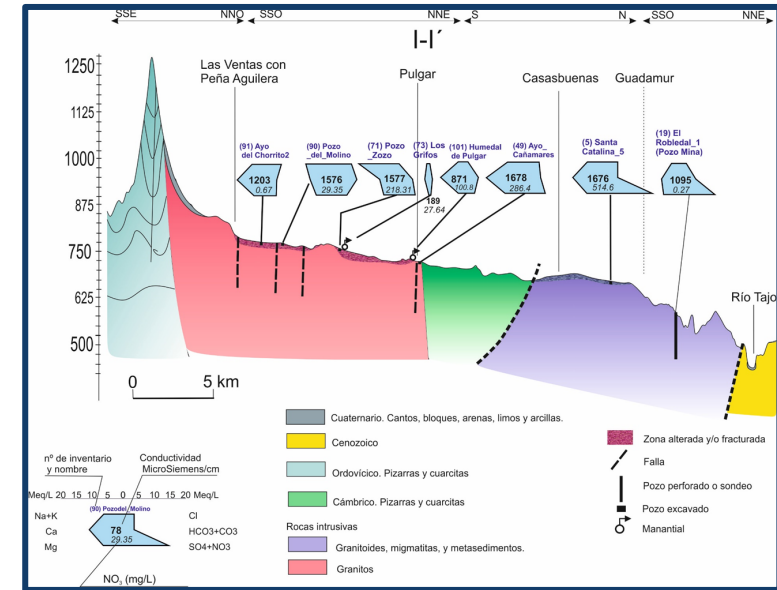
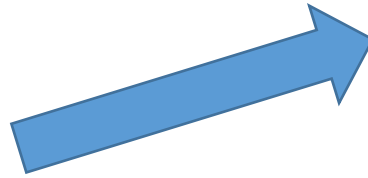


Source: Geologic and Mining Spanish Institute. GEODE Cartografía geológica digital continua a escala 1:50.000 (WMS)



Global goals

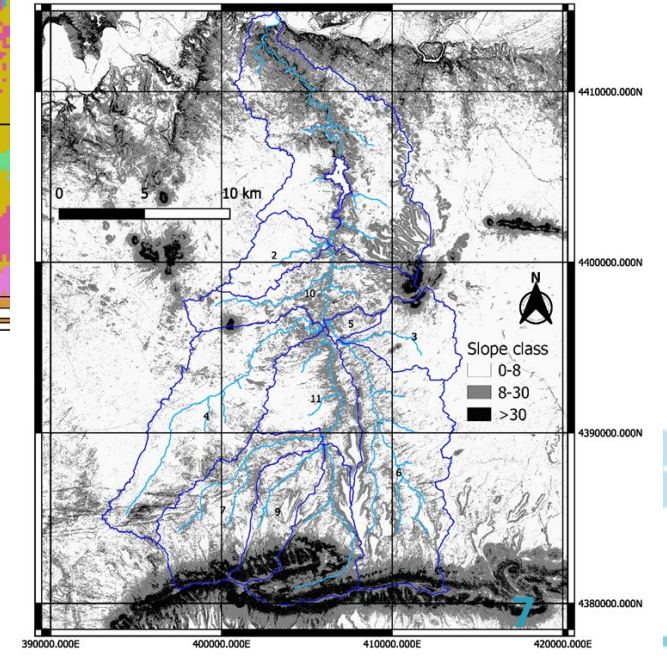
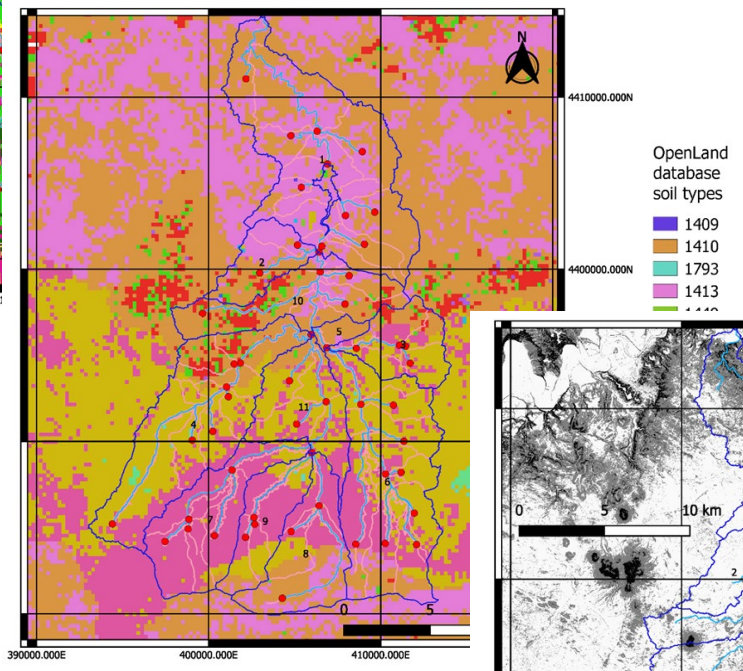
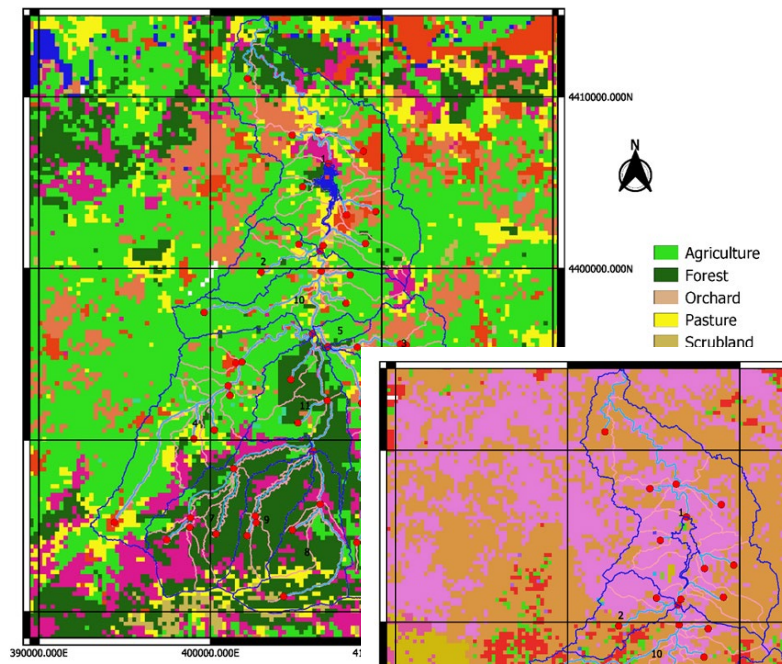
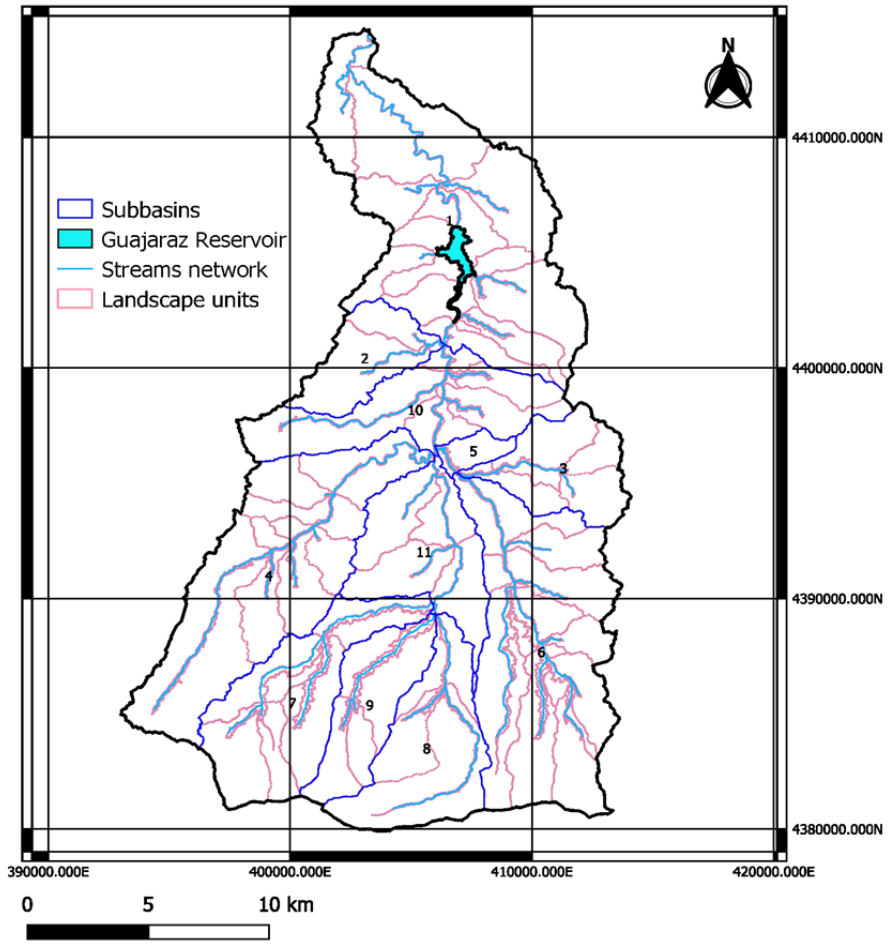
- Quantification of water resources in the Guajaraz River basin
- Hydrogeological characterization
- Hydrological modelling and climate change scenarios simulation



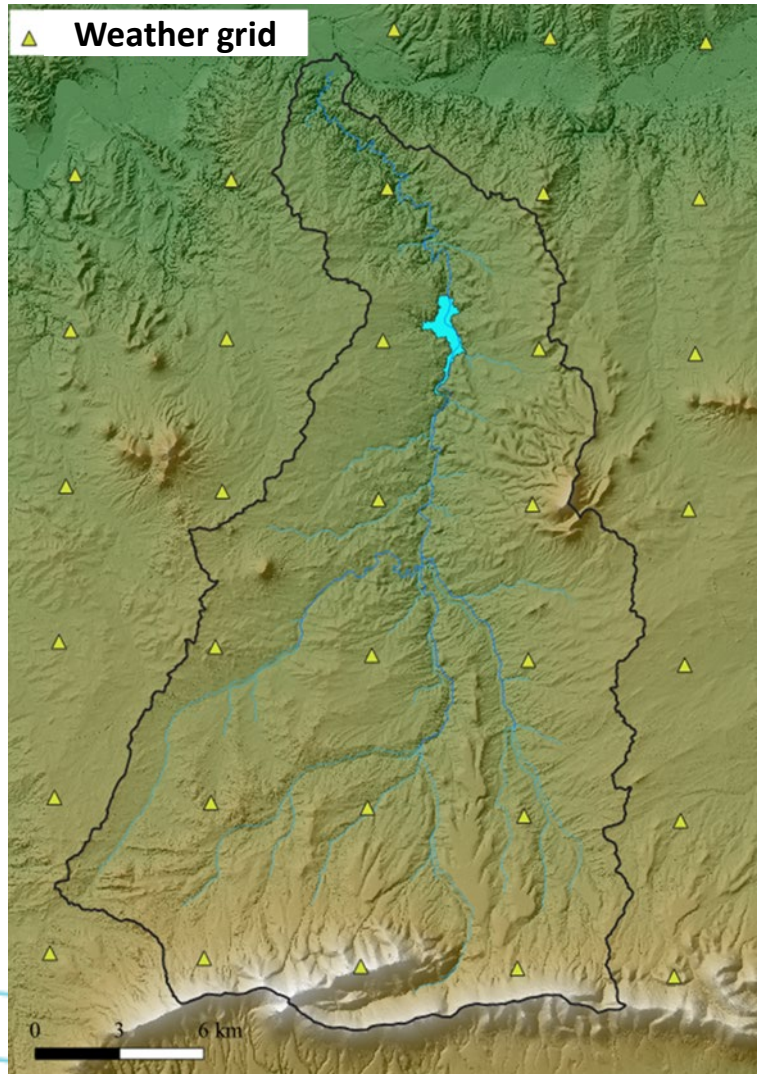
MODEL SET-UP AND DATA COLLECTION



Model set-up

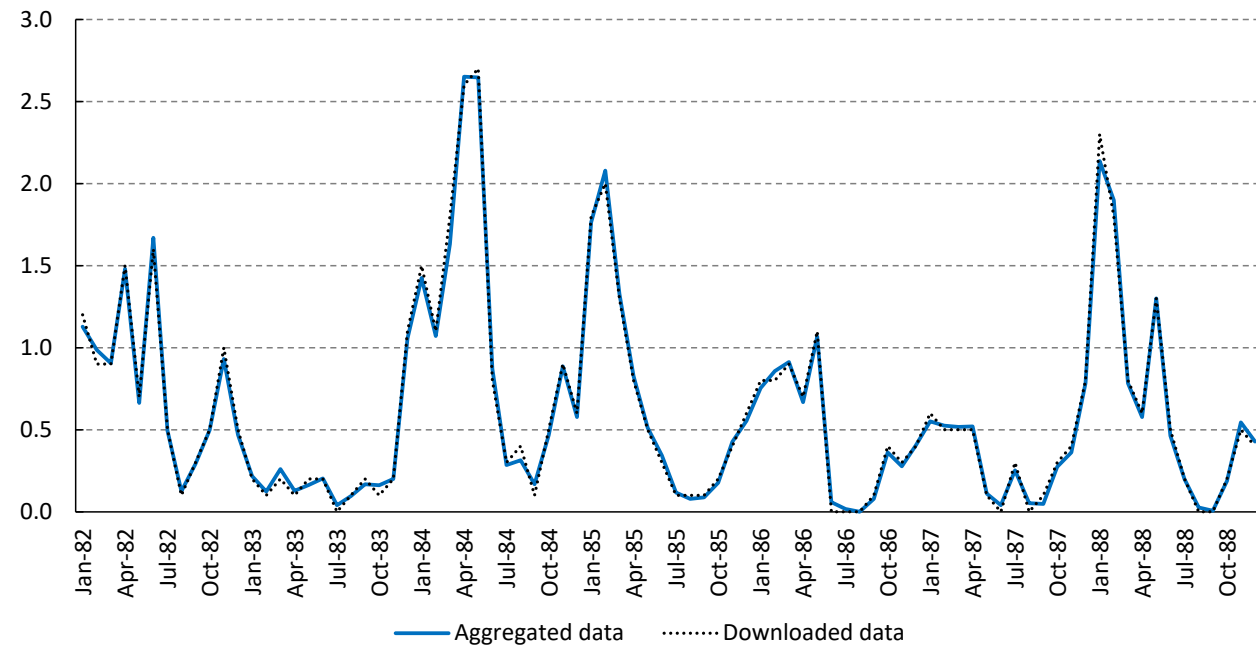


Data collection



- 5 x 5 km weather gridded data
(*Senent-Aparicio et al., 2021*)

- Monthly reservoir inflows



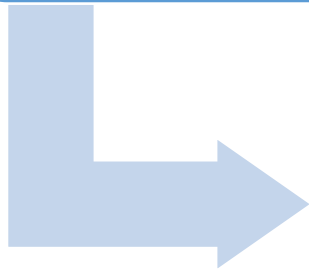
MODEL CALIBRATION



Initial workflow

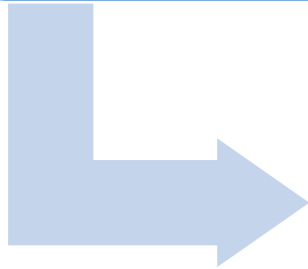
Sensitivity analysis

- Initial parameters selection



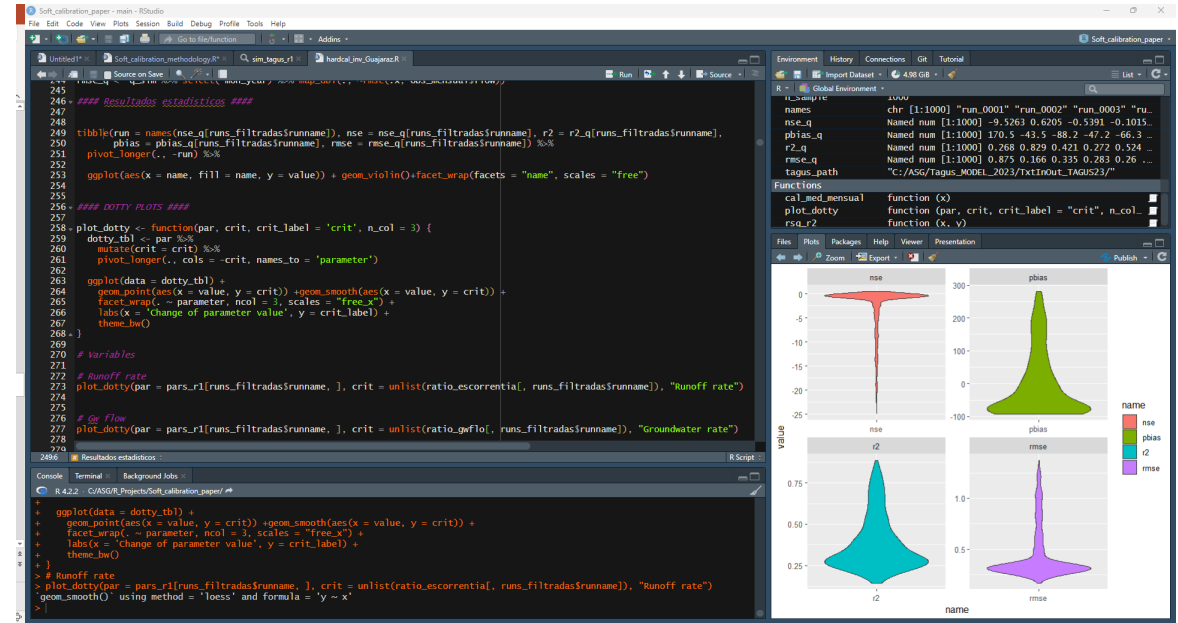
Soft calibration

- Parameter ranges restricted

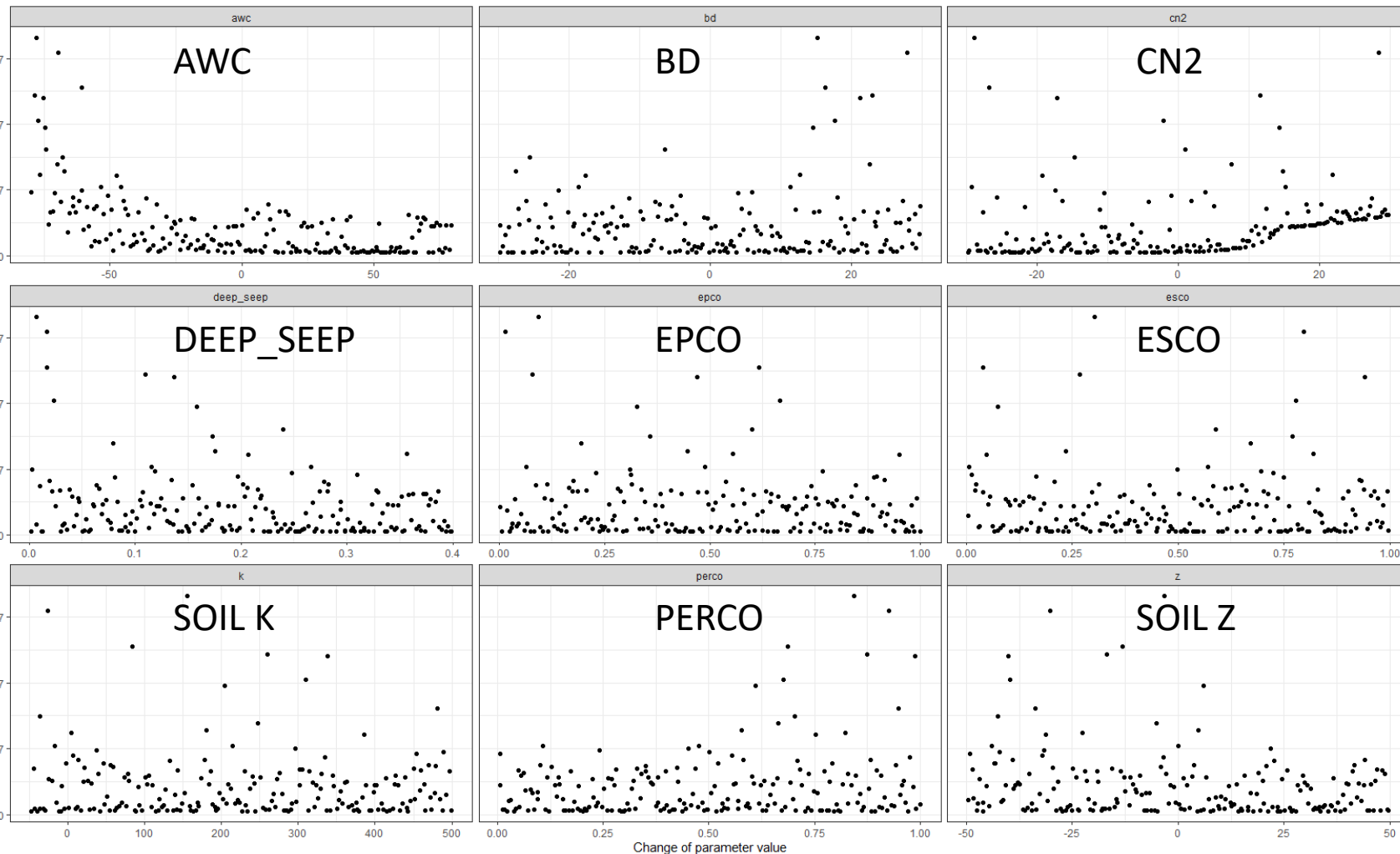


Hard calibration and validation

- New parameters added



Sensitivity analysis



Goal: parameter selection for soft calibration

Two iterations of 200 simulations. Sensitivity tested against several output variables, including reservoir inflow, ET and percolation.

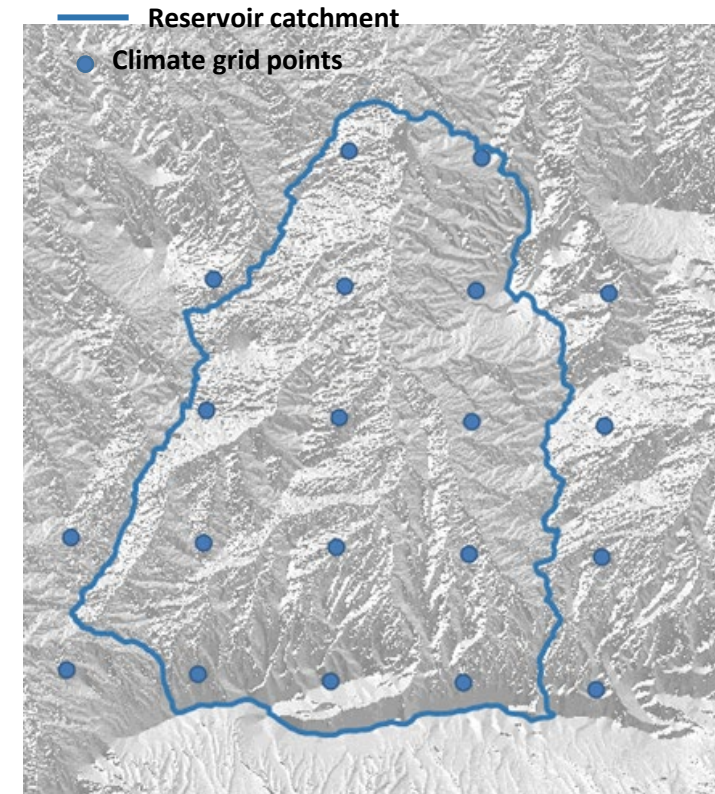
Lateral flow parameters less sensitive: Dismissed

Final parameter selection for soft calibration: 9 parameters

- **Extraction of soft data for calibration:**

- Runoff coefficient $\rightarrow \approx 4\%$
- Groundwater contribution to streamflow: estimated from neighbour catchments $\rightarrow \approx 20\%$

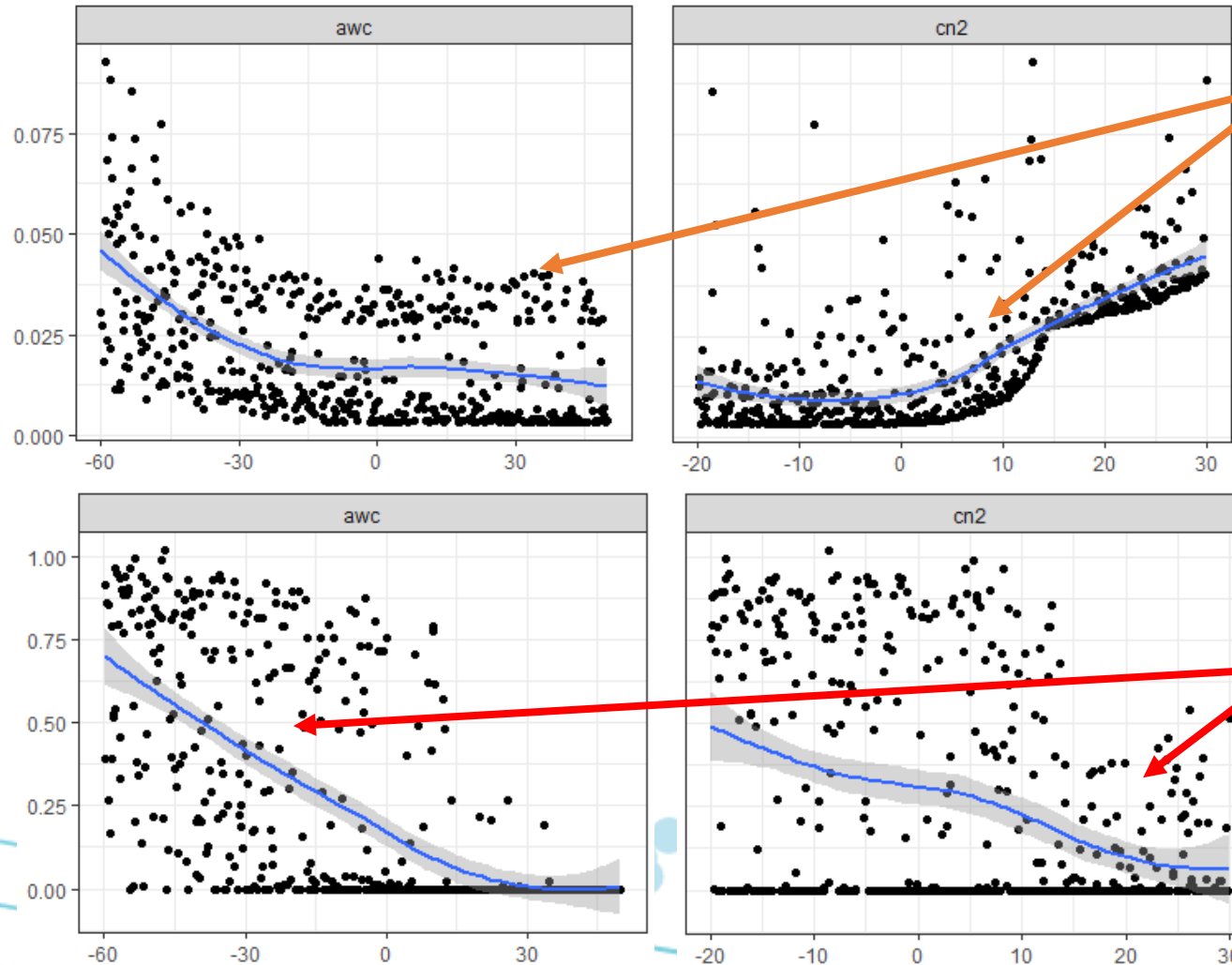
13:30 - 15:00	Session I2: Hydrology Mogens Zieler Stuen, Building 1422	Friday
13:30 - 13:50	Alejandro Sánchez Gómez Soft data collection for realistic hydrological modelling:	



- **Soft calibration:**

- **GOAL:** Restrict parameter values to guarantee realistic simulation in terms of water balance/streamflow components during subsequent hard calibration.
- 3 iterations of 500 simulations, constraining ranges after each iterations if trends

Example Round 2:



Sensitivity for Q coefficient \rightarrow Target \approx 4%

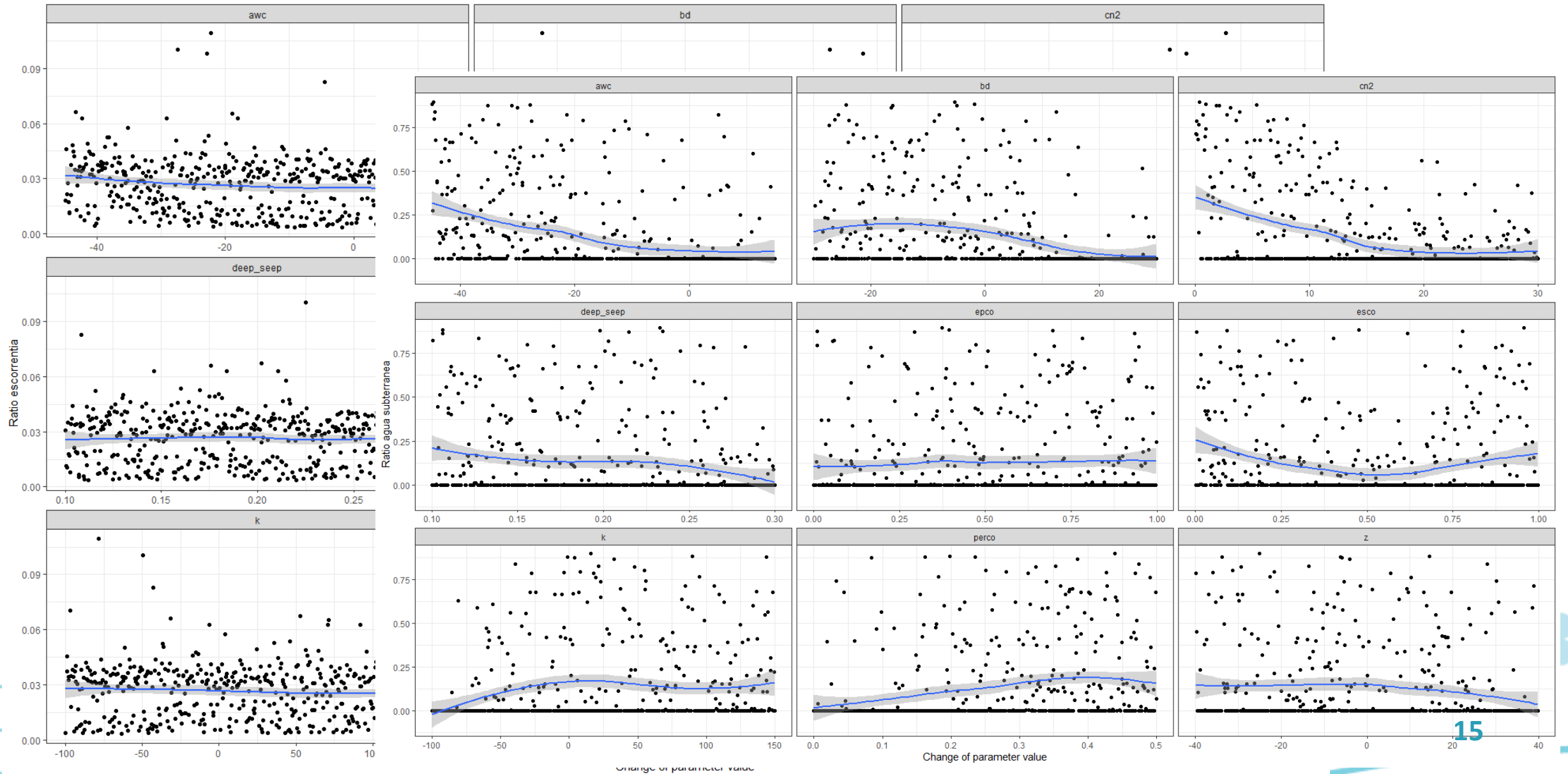
Contradictory = **challenge**

Sensitivity for GW ratio \rightarrow Target \approx 20%

Soft calibration

- Parameter restriction:

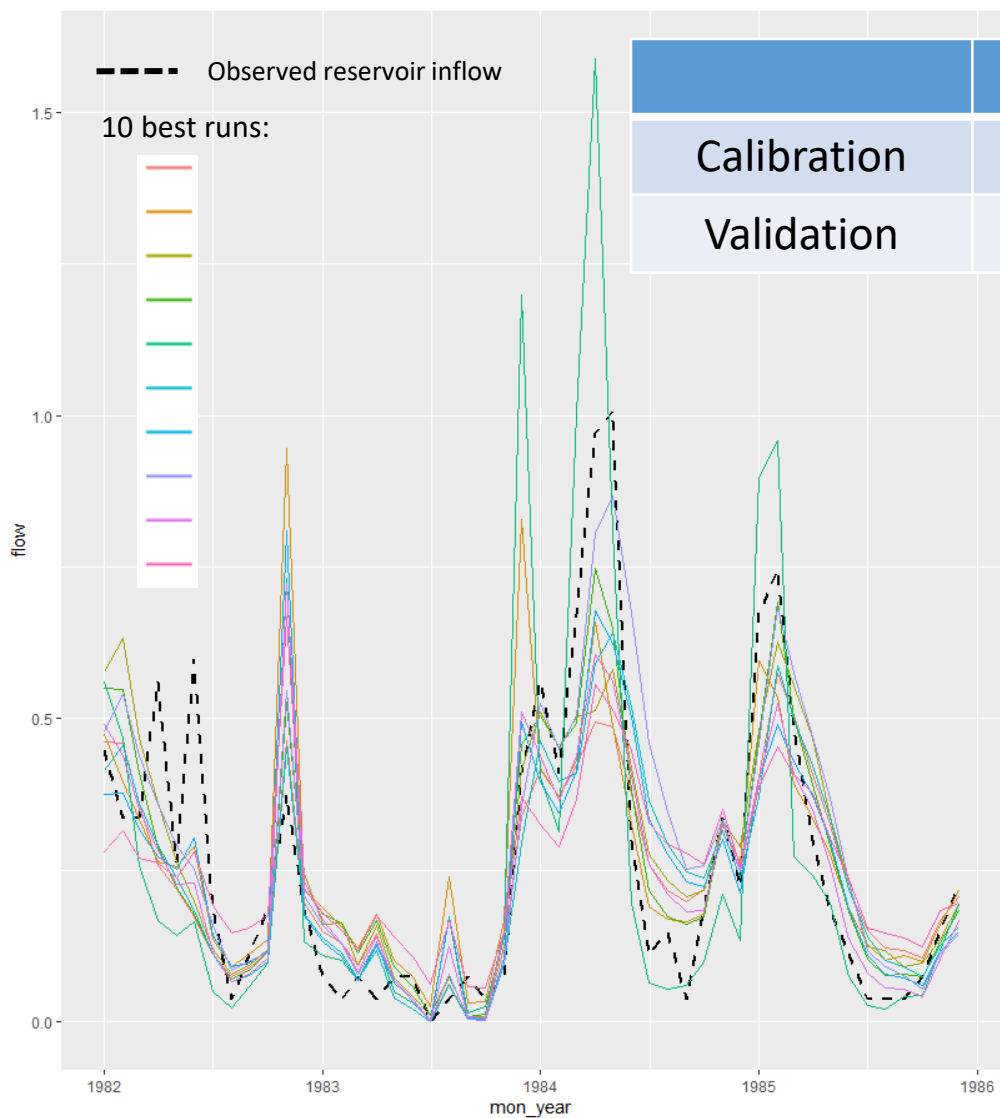
Parameter	Initial range	Final range	Change type
cn2.hru	-30, 30	0, 30	Percentage
perco.hru	0, 1	0, 0.5	Absolute value
epco.hru	0, 1	0, 1	Absolute value
esco.hru	0, 1	0, 1	Absolute value
z.sol	-50, 50	-40, 40	Percentage
k.sol	-100, 200	-100, 150	Percentage
bd.sol	-30, 30	-30, 30	Percentage
awc.sol	-80, 80	-45, 15	Percentage
deep_seep.aqu	0, 0.4	0.1, 0.3	Absolute value



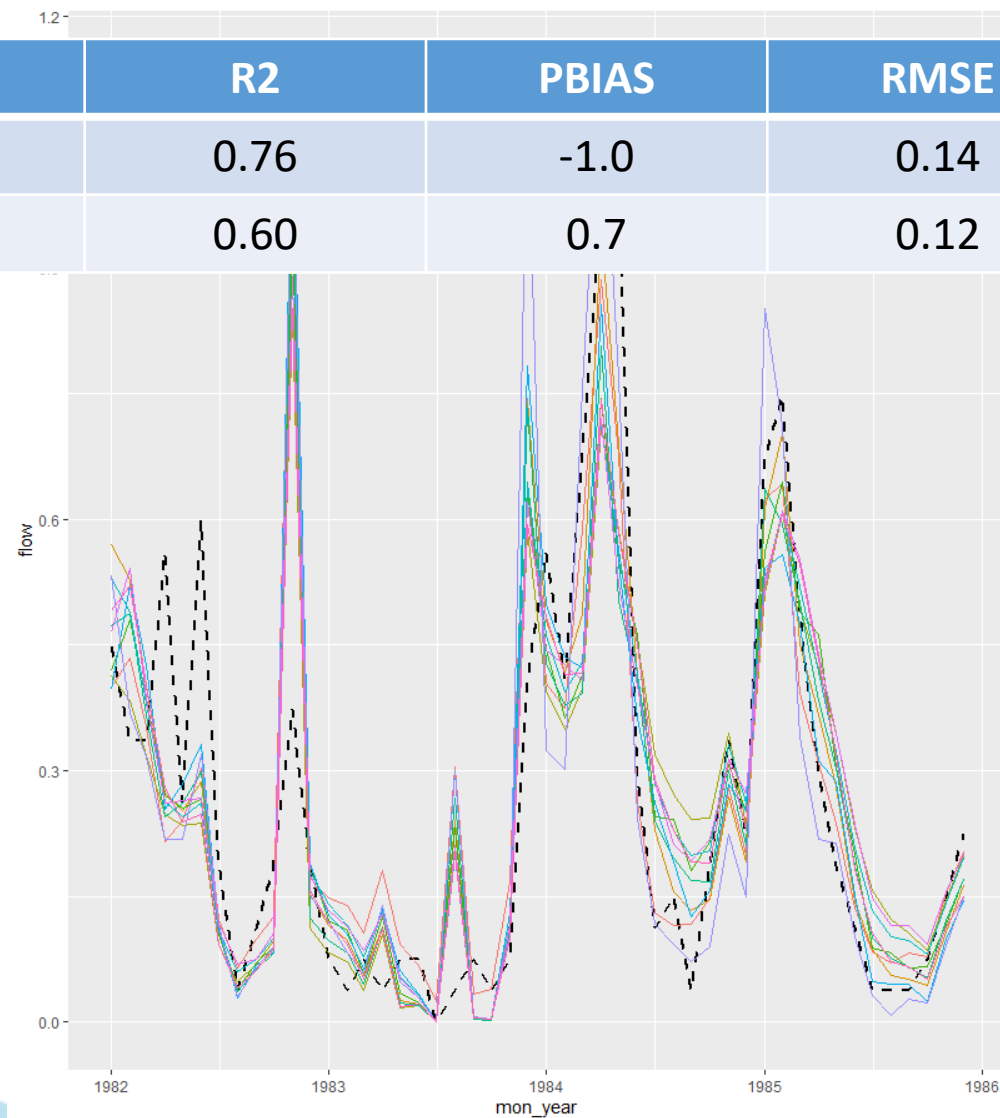
Hard calibration

- **GOAL:** To have a final version of the model which reproduces satisfactorily reservoir inflow in a realistic way
- 7 more parameters added based on our experience, related to lateral and groundwater flow.
- 3 additional iterations of 1000 simulations each
- Each round, restrict parameter values based on both statistics and soft data indices
- Calibration 82-85 , validation 86-88

Apparently, we can get satisfactory results from the very first iteration



	NSE	R2	PBIAS	RMSE
Calibration	0.75	0.76	-1.0	0.14
Validation	0.57	0.60	0.7	0.12

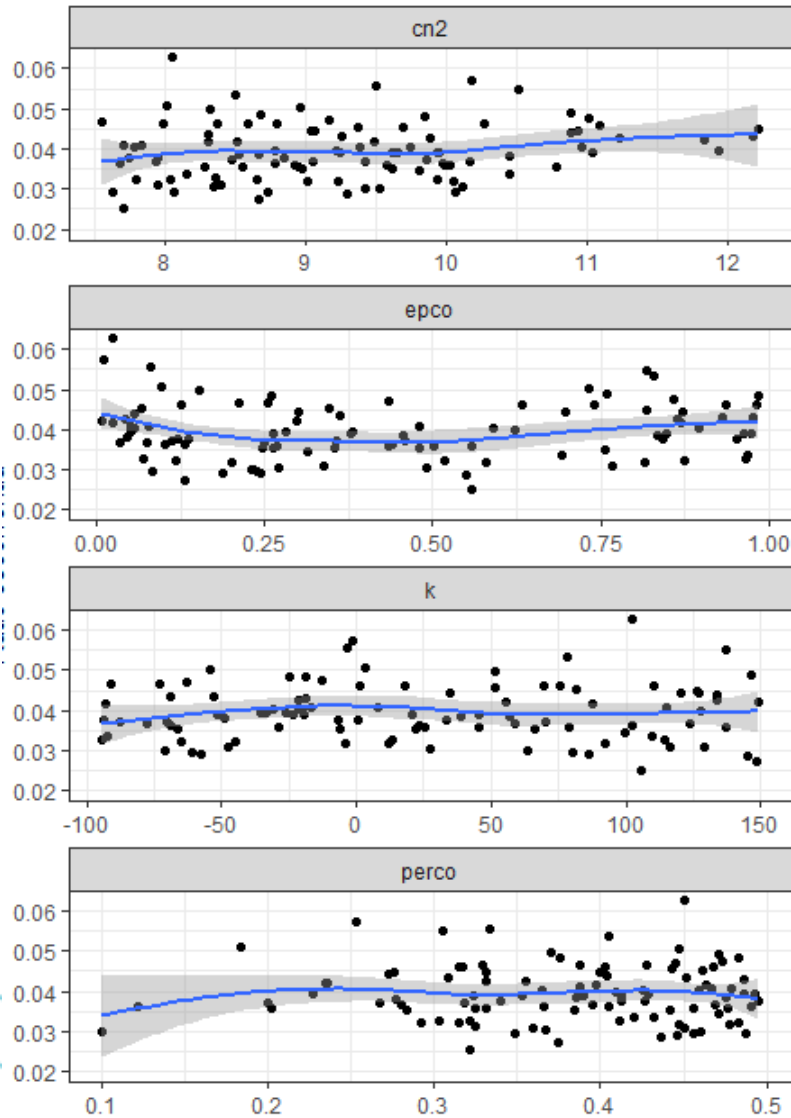


10 statistically-best simulations, 1st iteration

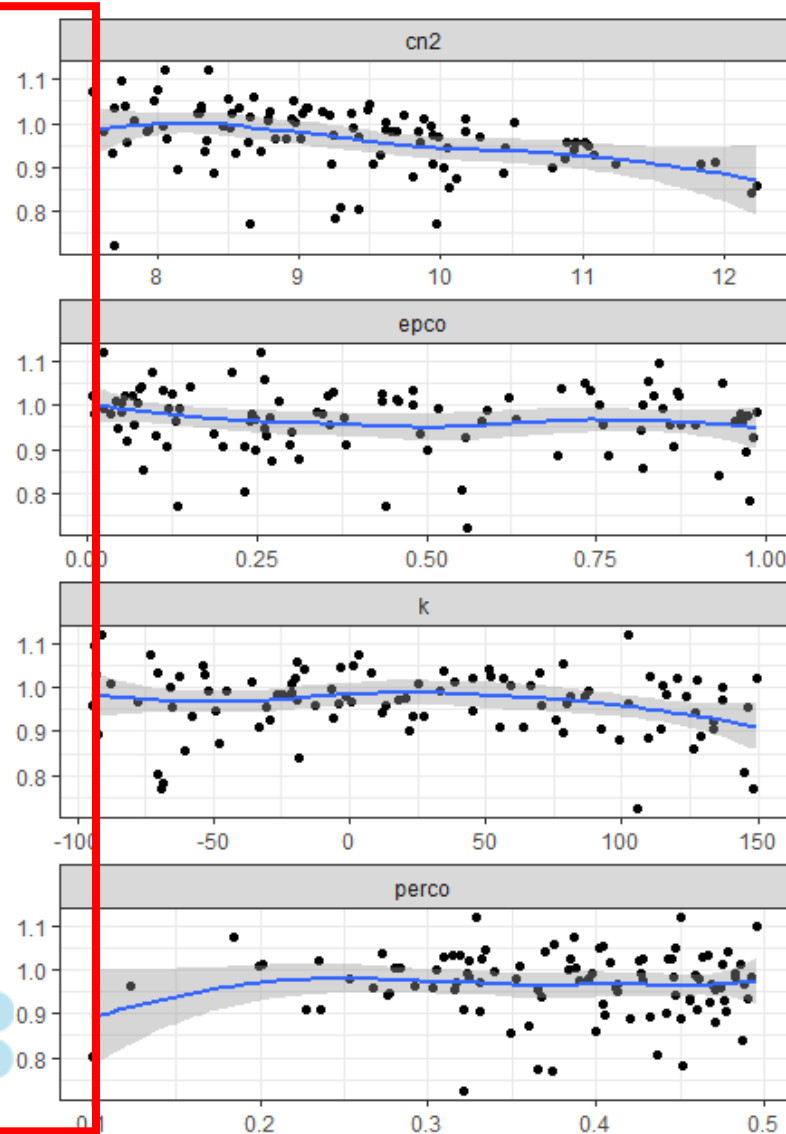
10 statistically-best simulations, 3rd iteration

HOWEVER:

Runoff ratio:



Groundwater index:

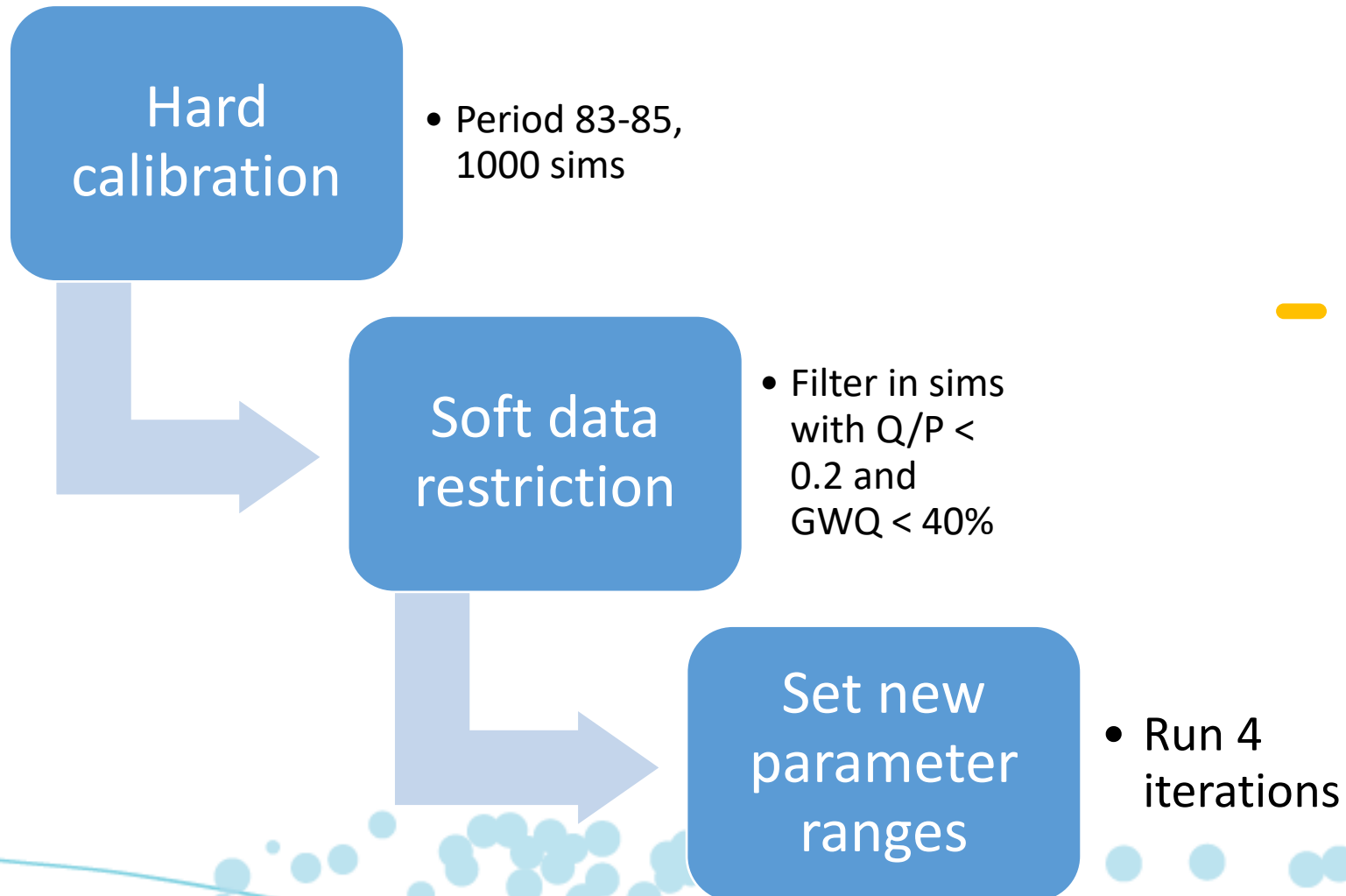


We filtered those with GW ratio lower than 0.5, but simulations were not accurate

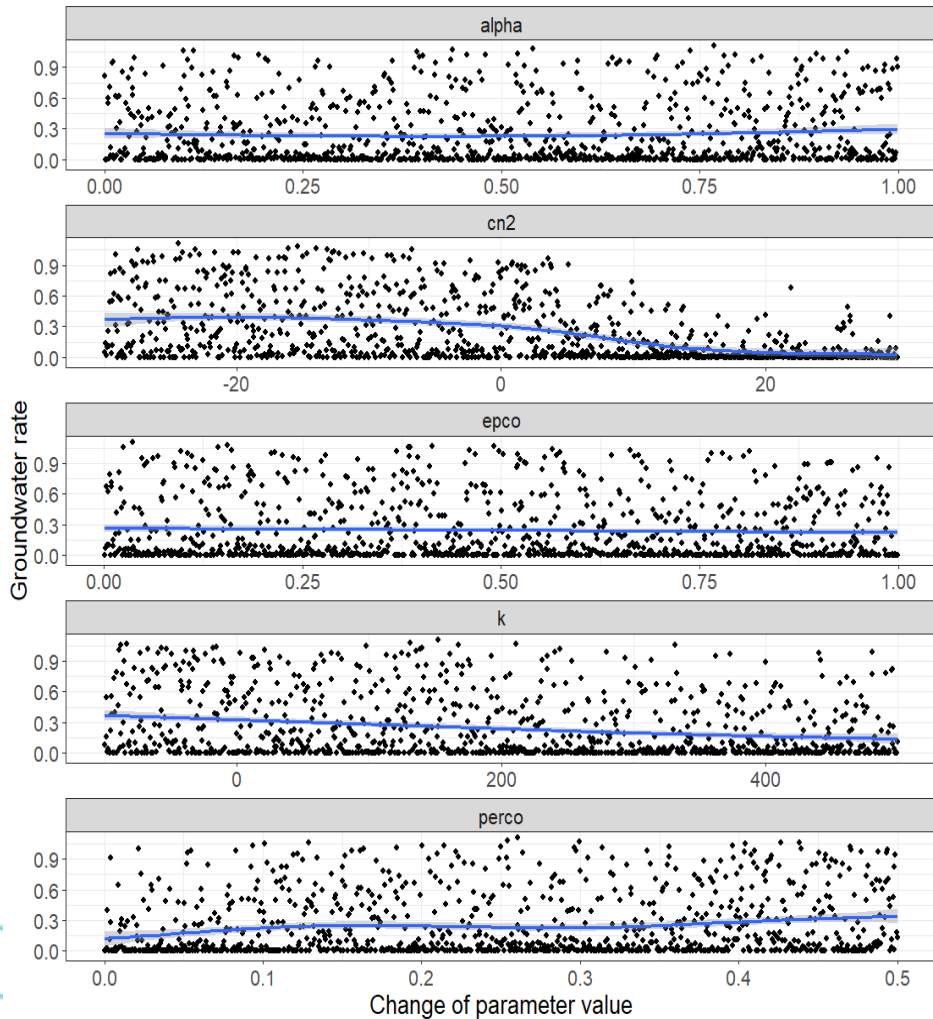
Assuming that the lack of data could be an issue, we tried an only-calibration scheme, but it did not work either

SOLUTION NEEDED to find another way to optimize model parameters

New workflow: INVERSE CALIBRATION

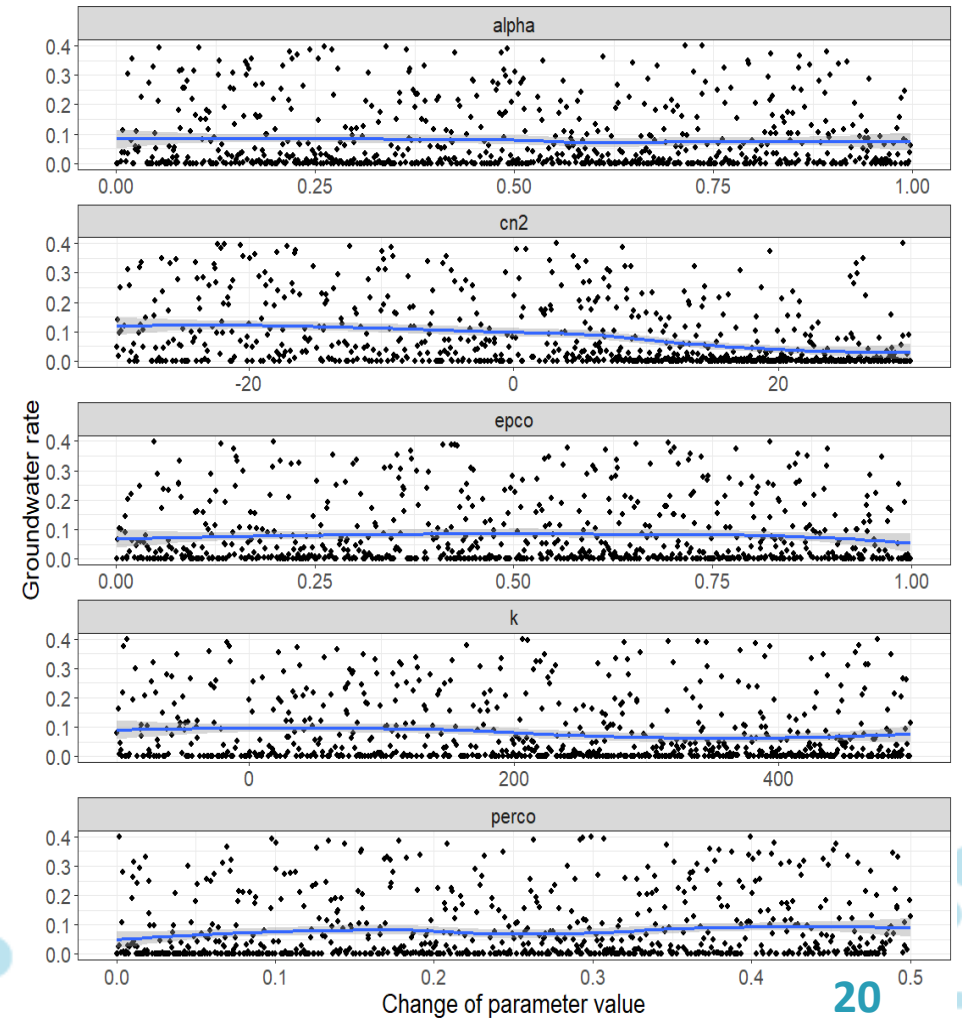


Soft data restriction after round 1 (filtering)



Without filter

With filter



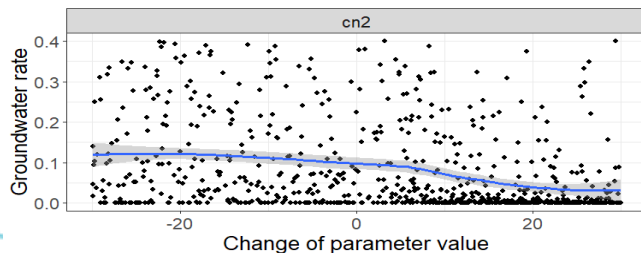
```

> rank_tableGRAN_r1
# A tibble: 719 x 12
  run      nse    r2 pbias  rmse rank_nse rank_r2 rank_rmse rank_pbias rank_run runoff_rate groundw_rate
  <chr>    <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
1 run_0257 0.611 0.791 -15.7 0.168   719   718   719   626   2782  0.0431  0.265
2 run_0681 0.420 0.530  5.9 0.205   715   654   715   675   2759  0.0574  0.237
3 run_0608 0.583 0.740 -30.3 0.174   718   712   718   556   2704  0.0368  0.385
4 run_0115 0.418 0.753 -36.7 0.206   714   715   714   529   2672  0.0320  0.327
5 run_0986 0.430 0.767 -40.1 0.204   716   717   716   522   2670  0.0308  0.181
6 run_0294 -0.0250 0.495  -6.8 0.273   676   633   676   670   2655  0.0461  0.313
7 run_0696 0.493 0.745 -48.9 0.192   717   714   717   500   2648  0.0244  0.379
8 run_0636 0.337 0.685 -43.8 0.220   712   702   712   514   2640  0.0278  0
9 run_0135 0.0696 0.419  7.3 0.260   694   583   694   667   2638  0.0596  0.204
10 run_0022 0.311 0.629 -43.2 0.224   710   694   710   515   2629  0.0283  0.239

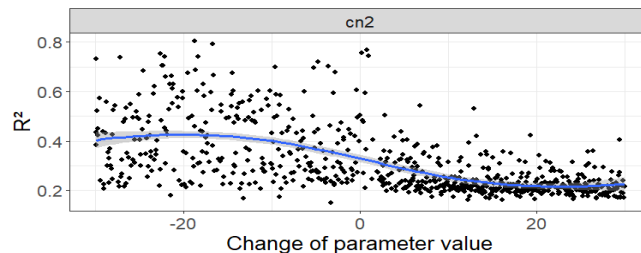
```

Selecting new parameter ranges

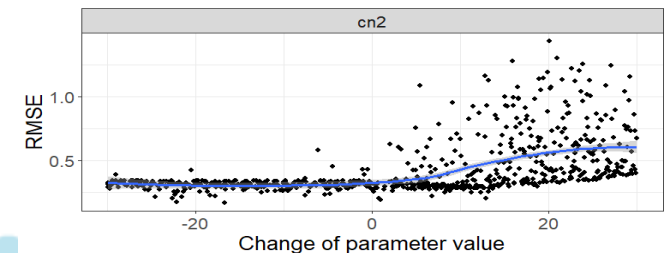
Groundwater rate



R²

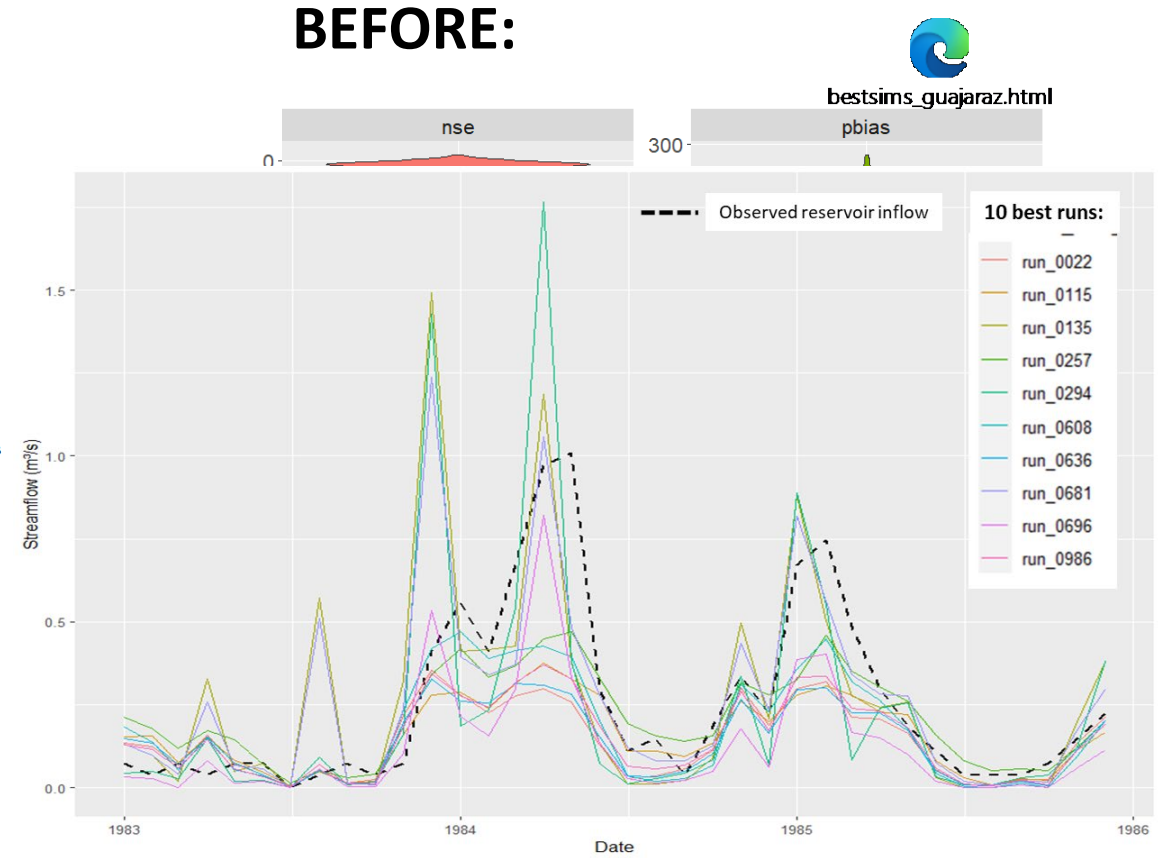
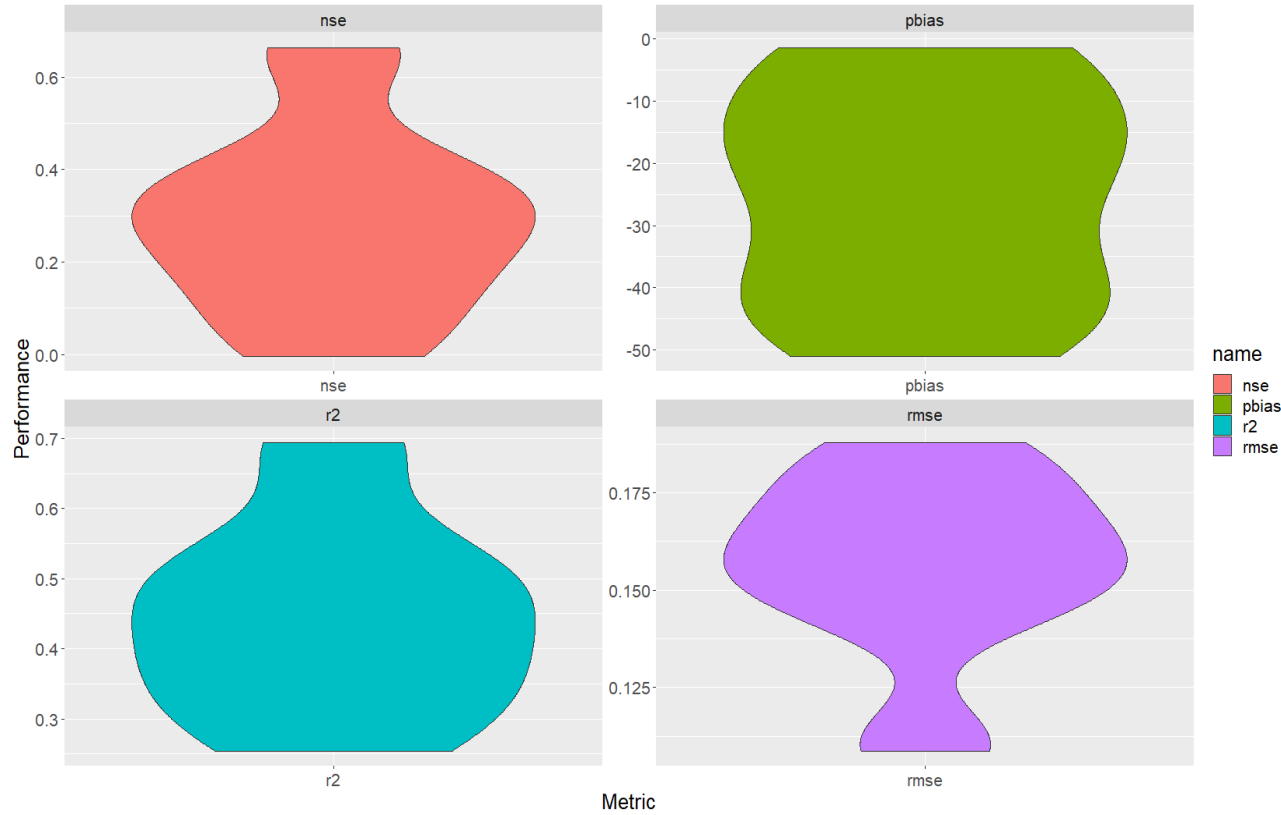


RMSE



Example for cn2, analysis repeated for the 17 parameters included

Performance overview

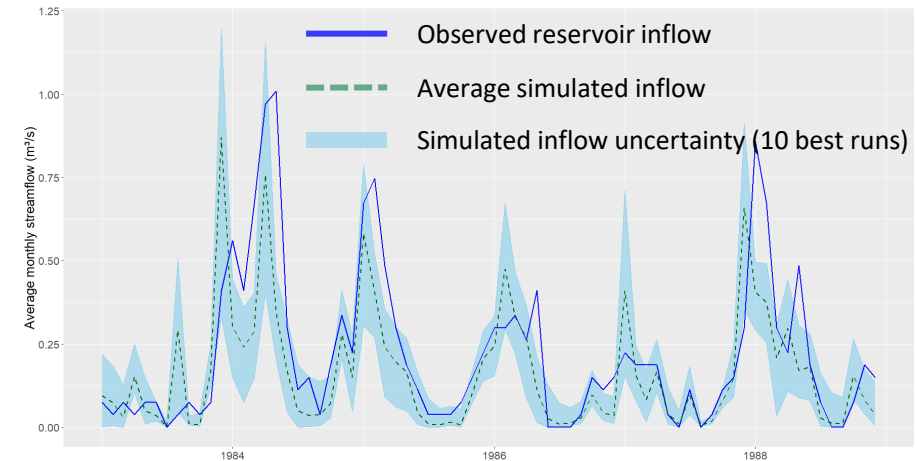


14 simulations selected:

Simulation	Round	NSE	R ²	PBIAS	RMSE	Q coef (%)	GW rate (%)
run_0257	R1	0.61	0.79	-15.7	0.168	4.31	26.5
run_0681	R1	0.42	0.53	5.9	0.205	5.74	23.7
run_0608	R1	0.58	0.74	-30.3	0.174	3.68	38.5
run_0309	R2	0.39	0.52	-37.4	0.21	3.4	5.13
run_0740	R2	0.32	0.45	-36	0.223	3.15	22.8
run_0440	R3	0.41	0.50	-26.8	0.208	4.08	34.3
run_0319	R3	0.16	0.54	1.6	0.248	5.35	35.3
run_0338	R3	0.41	0.51	-32.5	0.207	3.66	28.9
run_0655	R3	0.37	0.46	-15	0.215	4.72	39.9
run_1978	R4	0.46	0.57	-6.2	0.197	4.95	25.9
run_1245	R4	0.51	0.57	-17.1	0.188	4.58	31
run_1420	R4	0.30	0.51	-8.5	0.225	4.98	16.7
run_1616	R4	0.43	0.55	-20.2	0.203	4.37	39.6
run_0812	R4	0.59	0.69	-29.3	0.172	3.84	39.8

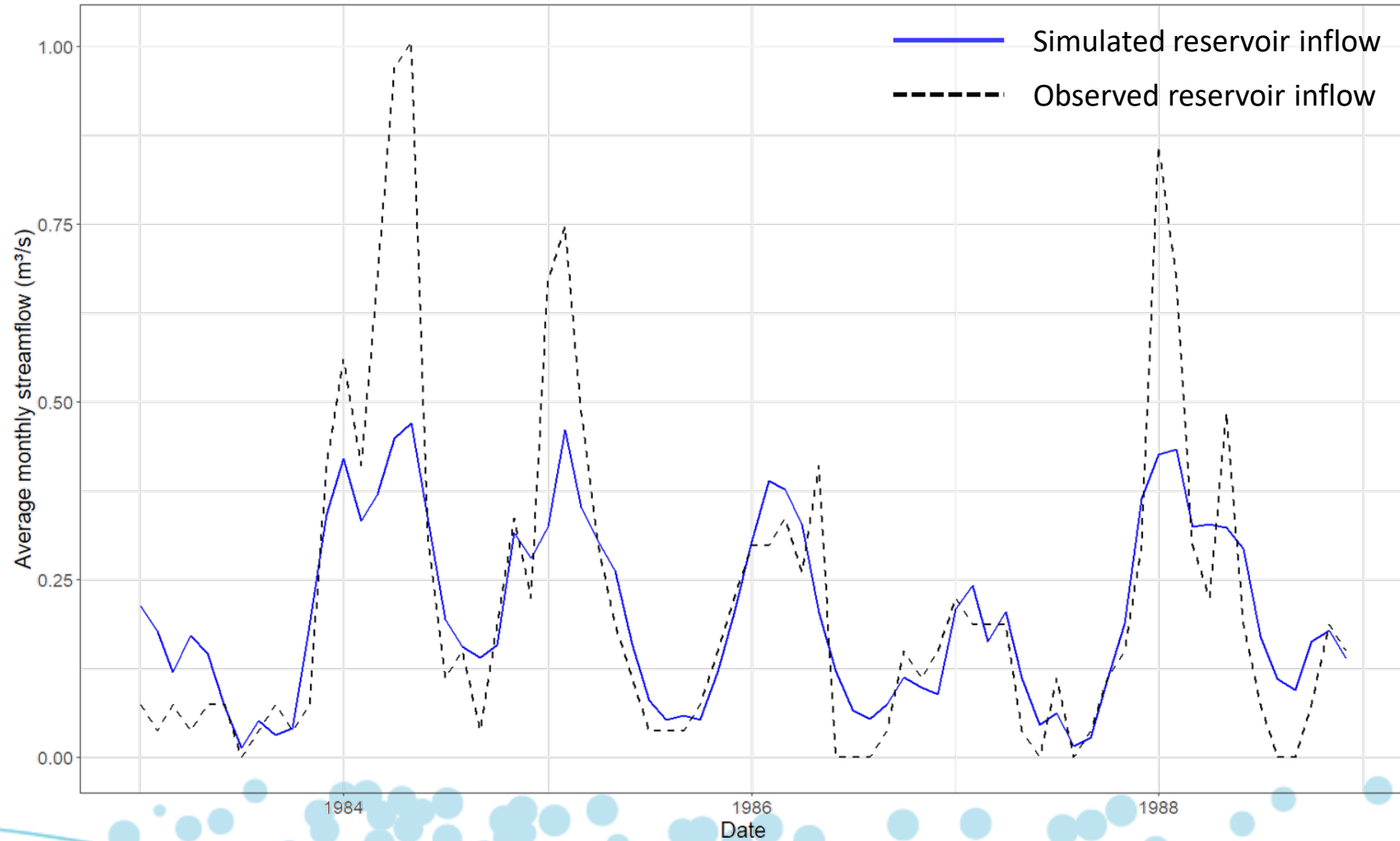
• Step 2: Validation + evaluation of selected simulations

- Validation period 86-88
- Assessment of performance metrics and WB indices
- Final simulation selection



CALIBRATION								VALIDATION					
Simulation	Round	NSE	R ²	PBIAS	RMSE	Q coef (%)	GW rate (%)	NSE	R ²	PBIAS	RMSE	Q coef (%)	GW rate (%)
run_0257	R1	0.61	0.79	-15.7	0.17	4.31	26.46	0.64	0.69	-1.50	0.11	3.81	26.90
run_0681	R1	0.42	0.53	5.9	0.21	5.74	23.70	0.17	0.34	-1.40	0.17	3.82	20.45
run_0608	R1	0.58	0.74	-30.3	0.17	3.68	38.49	0.66	0.69	-16.90	0.11	3.20	38.39
run_0309	R2	0.39	0.52	-37.4	0.21	3.40	5.13	0.16	0.37	-43.50	0.17	2.18	2.51
run_0740	R2	0.32	0.45	-36	0.22	3.15	22.82	0.30	0.53	-45.10	0.16	2.11	22.60
run_0440	R3	0.41	0.50	-26.8	0.21	4.08	34.30	0.00	0.25	-45.30	0.19	2.12	24.19
run_0319	R3	0.16	0.54	1.6	0.25	5.35	35.28	0.25	0.42	-21.40	0.16	3.03	40.80
run_0338	R3	0.41	0.51	-32.5	0.21	3.66	28.86	0.27	0.38	-32.50	0.16	2.61	33.93
run_0655	R3	0.37	0.46	-15	0.21	4.72	39.91	0.10	0.27	-24.30	0.18	2.93	47.30
run_1978	R4	0.46	0.57	-6.2	0.20	4.95	25.89	0.42	0.51	-3.90	0.14	3.71	33.59
run_1245	R4	0.51	0.57	-17.1	0.19	4.58	30.97	0.38	0.48	-12.30	0.15	3.39	37.97
run_1420	R4	0.30	0.51	-8.5	0.23	4.98	16.70	0.31	0.46	-15.00	0.16	3.28	20.72
run_1616	R4	0.43	0.55	-20.2	0.20	4.37	39.56	0.02	0.30	-51.00	0.19	1.90	24.33
run_0812	R4	0.59	0.69	-29.3	0.17	3.84	39.82	0.38	0.53	-37.00	0.15	2.43	43.82

Selected simulation hydrograph



CONCLUSIONS



- A SWAT+ model has been set up for the Guajaraz River basin
- Even using a soft calibration prior to a hard one, the statistically satisfactory simulations had a too high GW contribution to streamflow → could lead to unrealistic results
- A new calibration workflow was designed within SWATplusR, restricting simulations after hard calibration based on soft data
- Realistic simulations were obtained and still statistically satisfactory → They will guarantee a more accurate simulation of climate change scenarios

SUNTORY

THANK YOU!

