

Multi-objective calibration of a distributed eco-hydrological model using several remotely sensed information

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0.8 ≤ R < 1.0

0.6 ≤ R < 0.8

0.4 \$ R < 0.6

0.2 ≤ R < 0.4

0.0 ≤ R < 0.

.02 < R < 0.0

-0.2 ≤ R < -1.0

0.8 ≤ R < 1.0

0.6 ≤ R < 0.8

0.4 ≤ R < 0.6

0.2 ≤ R < 0.

0.0 ≤ R < 0.2

-0.2 < R < 0.0 -0.2 ≤ R < -1)

MINISTERIO DE ECONOMIA Y COMPETITIVIDAD

INTRODUCTION

Calibration of eco-hydrological models is difficult to carry on, even more if observed data sets are scarce. The traditional calibration approach mainly focuses on the temporal variation of the discharge at the catchment outlet point, representing an integrated catchment response and provides thus only limited insight on the lumped behaviour of the catchment. It has been long demonstrated the limited capabilities of such an approach when models are validated at interior points of a river basin.

The development of distributed eco-hydrological models and the burst of spatio-temporal data provided by remote sensing appear as key alternative to overcome those limitations. Indeed, remote sensing imagery provides not only temporal information but also valuable information on spatial patterns, which can facilitate a spatial-pattern-oriented model calibration

It is still unclear whether including spatio temporal data improves model performance in face to an unavoidable more complex and time-demanding calibration procedure. To elucidate in this sense, we performed three different multiobjective calibration configurations: (1) including only temporal information of discharges at the catchment outlet (2) including both temporal and spatio-temporal information and (3) only including spatio-temporal information. In the three approaches, we calibrated the same distributed eco-hydrological model (TETIS) in the same study area: Rambla de la Viuda catchment, and used the same multi objective algorithm: MOSCEM-UA. The spatio-temporal information obtained from satellite has been the surface soil moisture (from SMOS-BEC) and the leaf area index (from MODIS)





METHODOLOGY

Configuration 1

Main state variable: flow at the catchment outlet (Q) Mono-objective calibration: Nash-Sutcliffe Efficiency index (NSE)

Configuration 2

Main state variables: flow at the catchment outlet (Q) and near-surface remotely sensed soil moisture (SM) Multi-objective calibration: Nash-Sutcliffe Efficiency (NSE) and Spatio-Temporal Efficiency (STE)

Configuration 3: UNGAUGED BASIN

Main state variables: remotely sensed leaf area index(LAI) and near-surface remotely sensed soil moisture (SM) Multi-objective calibration: Spatio-Temporal Efficiency (STE) and Spatio-Temporal Efficiency (STE)



Q^t_{sim} is modelled discharge at time t Q^t_{abs} is observed discharge at time t, $\overline{Q_{obs}}$ is the mean of observed discharges



1) Multi-objective approaches (configurations two and three) lead to better model performance. The visualization of the Pareto set allows to identify whether all objective functions can be simultaneously optimized and pinpoints the optimal set of parameters, assisting thus the parameter selection.

2) The differences among the configurations are more evident during the validation period, pointing out which are (configuration two and three) more robust and which lead to more realistic effective parameter values

3) Even though challenging, spatio-temporal data, in particular near-surface SM and LAI, can be used as relevant source of information to calibrate process-based models.

4) The SMOS/MODIS fine-scale surface soil moisture and MODIS LAL both remotely sensed data are consistent to carry out calibration process, even more in combination to improve robustness

5) It is possible to use only remotely sensed information (i.e. configuration three) to calibrate satisfactorily a distributed eco-hydrological model (i.e. TETIS)

ACKNOWLEDGEMENTS AND REFERENCES

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95% of the variance

the principal component i

component i

 $STE = \sum w_i \cdot NSE(load_i)$

N = number of principal components that explains at least

 w_i = portion of explained variance in the principal

load_i = loadings of the observed and simulated data in





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