





Presentation of the fuzzy expert systems, possible use in the CN-MIP project

Gianni BELLOCCHI

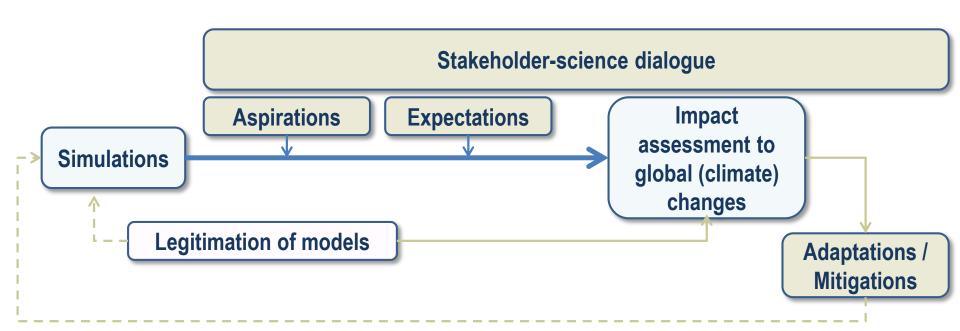
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"We conserve many things that we don't evaluate and little of those we value" (Geoffrey M. Heal)

Deliberative process in model-based climate change studies



Bellocchi et al., 2006, Ital. J. Agrometeorol.

Rivington et al., 2007, Environ. Modell. Softw.

Bellocchi et al., 2014, Agron. Sustain. Dev.

Institutionalising deliberative practices for context-specific model evaluations

Model evaluations are (sometimes) an (important) orientating landmark in the skyline of decisions, without replacing them

To evaluate (crop and grassland) simulation models is far more urgent as many of the (tactical and strategic) decisions (in agriculture) are based on model outcomes

Dealing with (existing) and designing (new) agricultural systems is a priority that deliberations about model evaluation contribute to accomplish in a more efficient (maybe more appropriate) manner, in any case with more awareness if (genuine) collective deliberations are possible

The central issue is to think and conceive model evaluation in a (clear) decisional perspective about type of model, operability, transparency, etc.

As several models are at hand, "mod-diversity" imposes the analysis of case-by-case issues, while also integrating the specific context in a larger-scale perspective (in space and time)

A Review of Methodologies to Evaluate Agroecosystem Simulation Models

F. MARTORANA and G. BELLOCCHI

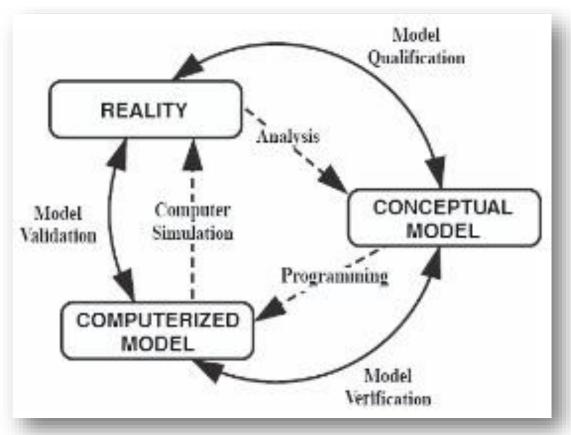


Katja Richter Clement Atzberger Tobias B. Hank Wolfram Mauser An Indicator of Solar Radiation Model Performance based on a Fuzzy Expert System

Gianni Bellocchi,* Marco Acutis, Gianni Fila, and Marcello Donatelli

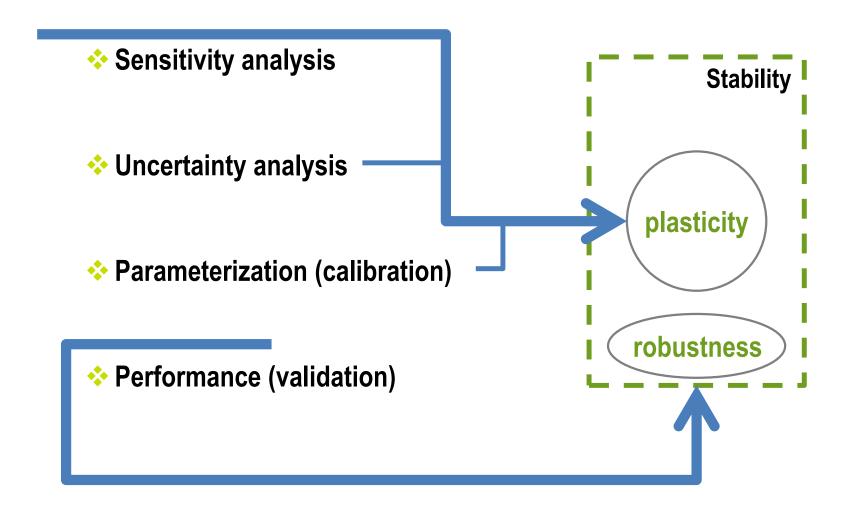
Model evaluation

Model evaluation (validation + verification): action in which the quality of a mathematical model for specific objectives is established



Schlesinger, 1979, Simulation

Model evaluation components



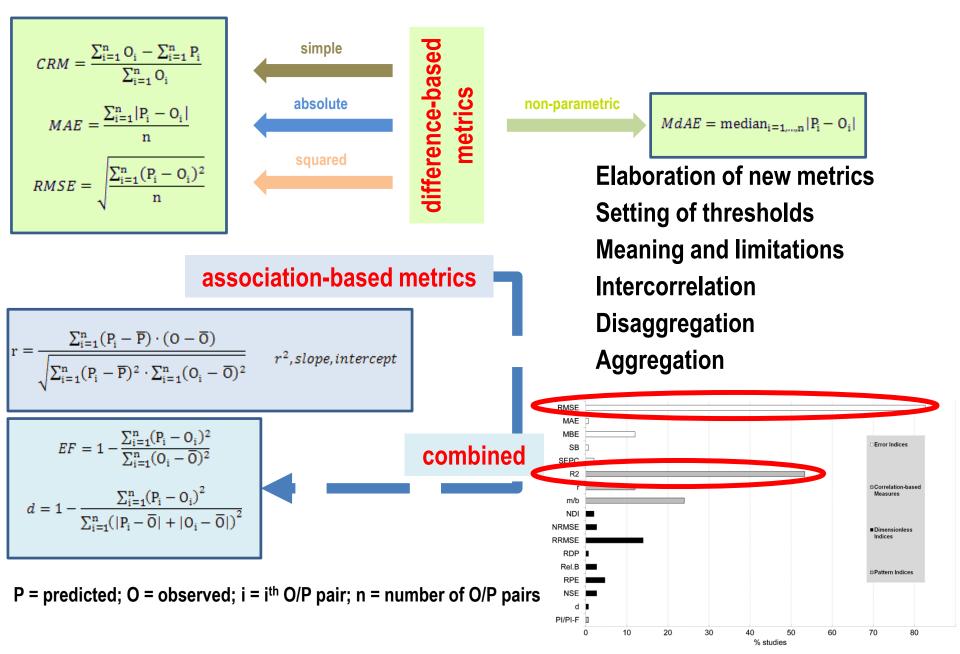
Key issues

Vov validation issues	Major factors to investigate					
Key validation issues	Modelling	Model	Model	Model	Modelling	
	objective	inputs	outputs	structure	d onditions	
Validation purpose	X		X		X	
Robustness of results			X	l	X	
Interpretation of		X	X	X	1	
phenomena — — —					<u>.</u>	
Model comparison				X	l i	
Model predictions -	x		-X-		- X -	
Model complexity		X	X	X	1	
Data accuracy		X	X		i	
Time histories			X		i	

Bellocchi et al., 2010, Agron. Sustain. Dev.

Approaches to model performance evaluation

- Indices
 - difference-based indices (RMSE, CRM, MAE, ...)
 - * association-based indices (slope, intercept, r, r², EF, ...)
 - pattern indices (range-based, F-based)
- Test statistics
 - Student-t, chi square, ...
- Probability distributions
 - cumulative, density functions
- Time mismatch analysis
- Aggregation of statistics
 - hierarchical, fuzzy logic based indicators



Richter et al., 2012, J. Appl. Remote Sens.

Setting of thresholds

Performance measure	Unit	Value range and purpose	Reliability criteria
Coefficient of determination (R ²) of the linear regression estimates versus measurements	dimensionless	0 (absence of fit) to 1 (perfect fit): the closer values are to 1, the better the model	> 0.8
Willmott (1982) index of agreement (d)	dimensionless	0 (absence of agreement) to 1 (perfect agreement): the closer values are to 1, the better the model	> 0.8
Mean absolute error over the mean of the measured values (MAE(%))	%	0 (optimum) to positive infinity: the smaller MAE(%), the better the model performance	< 20

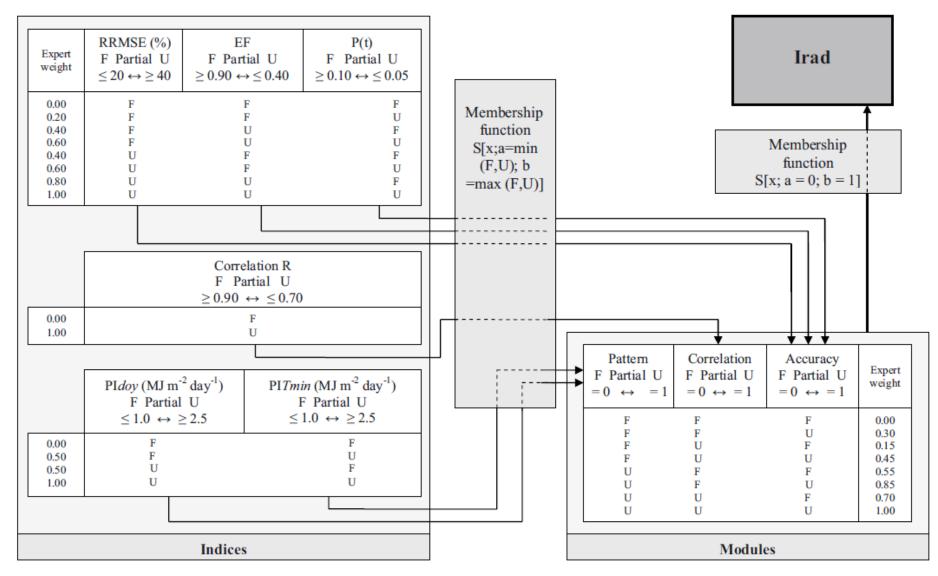
Properties of evaluation metrics

- Non-dimensionality
 - to avoid influences from the magnitude of the values
- Lower and upper bounding
 - for easy comprehension (e.g. between 0-no agreement and 1-full agreement)
- Symmetry
 - datasets interchangeability (e.g. Reduced Major Axis instead of Least Squares regression)
- Same unit of data (difference-based metrics)
 - to understand the magnitude of the error
- Prediction capability
 - comparison to measurement statistics (e.g. efficiency with respect to observed mean value)

Latest developments

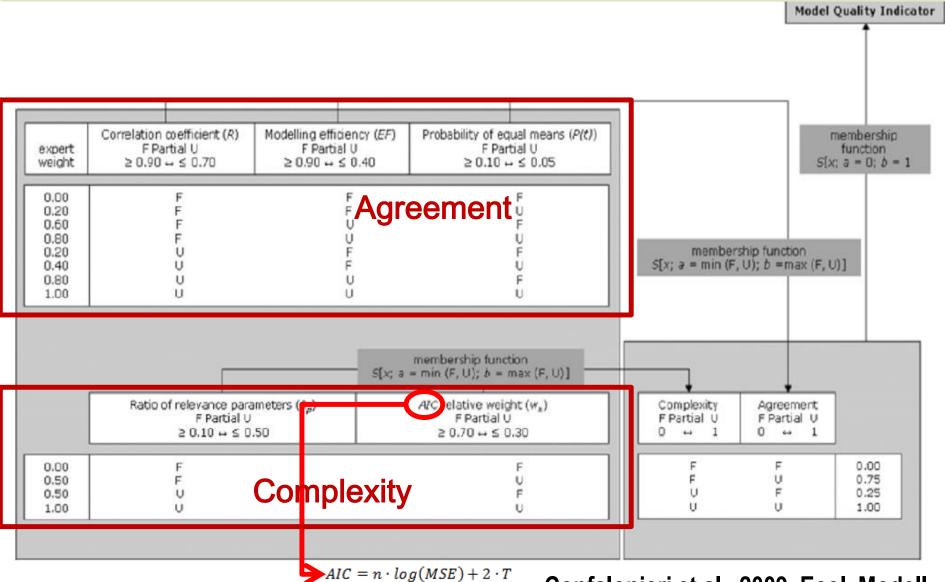
- ❖ Aggregation of multiple evaluation metrics into integrated indicators (fuzzy logic principle, after Bellocchi et al., 2002, Agron. J.)
- Inclusion, in the evaluation of models, of other measures than performance metrics, such as sensitivity analysis measures and information criteria for model selection (Confalonieri et al., 2009, Ecol. Modell.)
- Elaboration of the model robustness concept (Confalonieri et al., 2010, Ecol. Modell.)
- Elaboration of the model plasticity concept (Confalonieri et al., 2012, Ecol. Modell.)

Fuzzy logic-based integrated indicators / 1

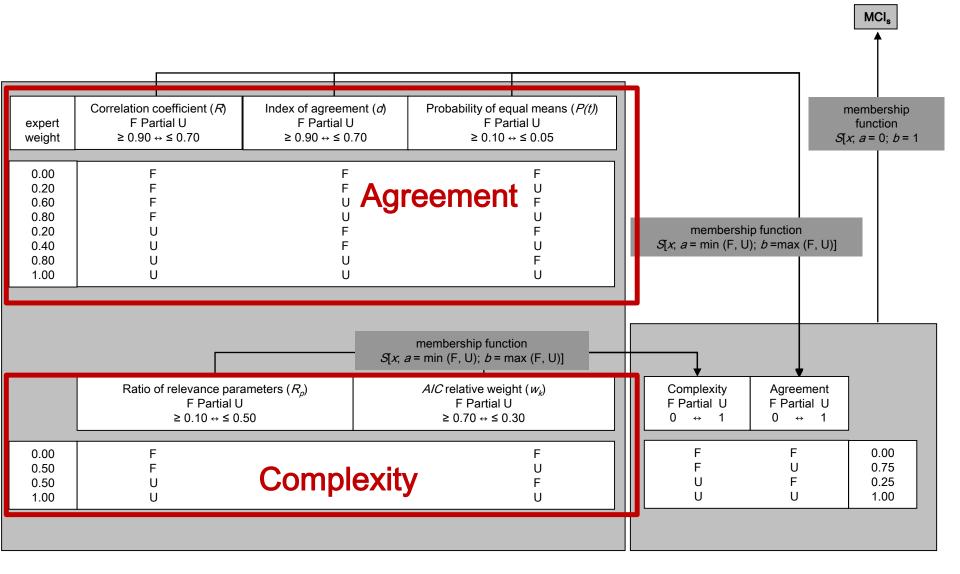


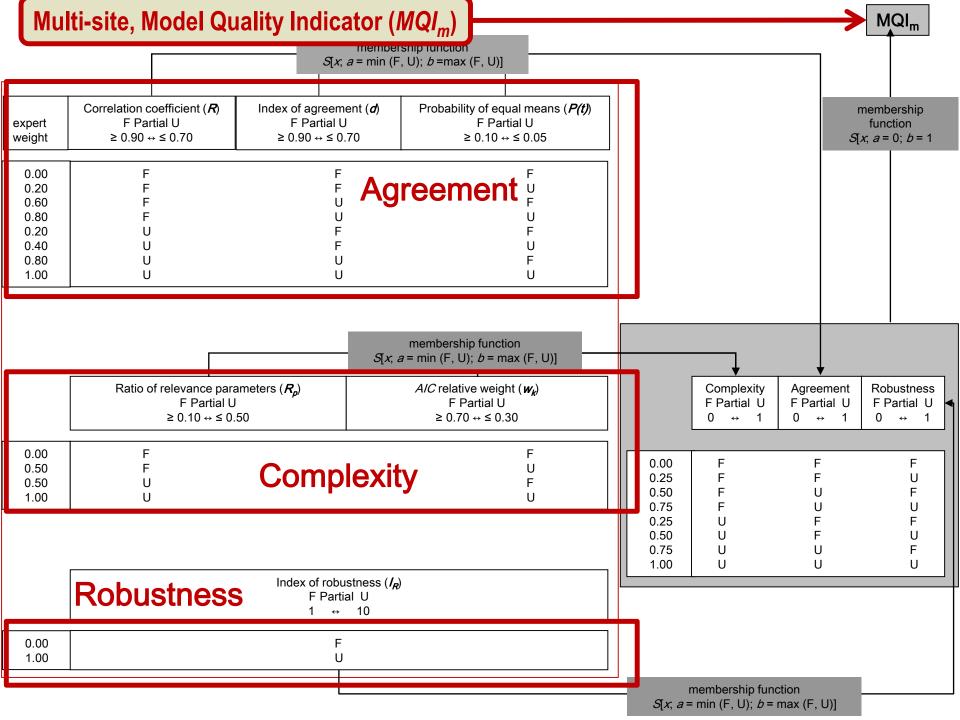
Bellocchi et al., 2002, Agron. J. Rivington et al., 2007, Agr. Forest Meteorol.

Fuzzy logic-based integrated indicators / 2

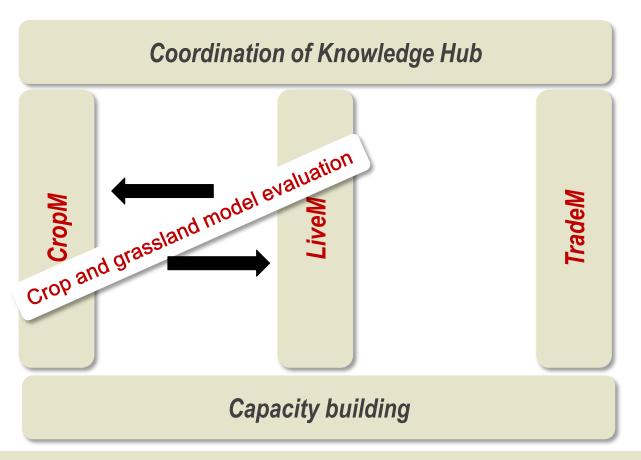


Confalonieri et al., 2009, Ecol. Modell.





MACSUR cross-cutting activities



CropM-LiveM

- Definition of model performance indicators
- Elaboration of model evaluation protocols

Model evaluation / deliberative process

rehensive evaluation Components of model quality

Agreement with actual data (rmetrics, test statistics)

Complexity (set of equations, parameters)

Stability (performance over different conditions)

Evaluation - simulation models

(experimental / observational research, socio-economic / climate scenarios)

Deliberative process

(review, exchange of information, consensus)

Context

Credibility

Transparency

Uncertainty

Background

Stakeholders

Bellocchi et al., 2014, Agron. Sustain. Dev

Synthetic indicators

Aggregation rules: fuzzy-logic based weighing system

Non-dimensionality

Lower and upper bounding

I. Agreement

- Correlation coefficient
- Index of agreement
- Probability of equal means

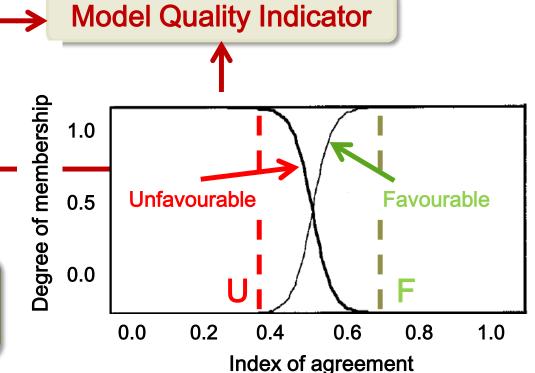
II. Complexity

- Ratio of relevant parameters
- Parameters-agreement criterion

III. Stability (robustness)

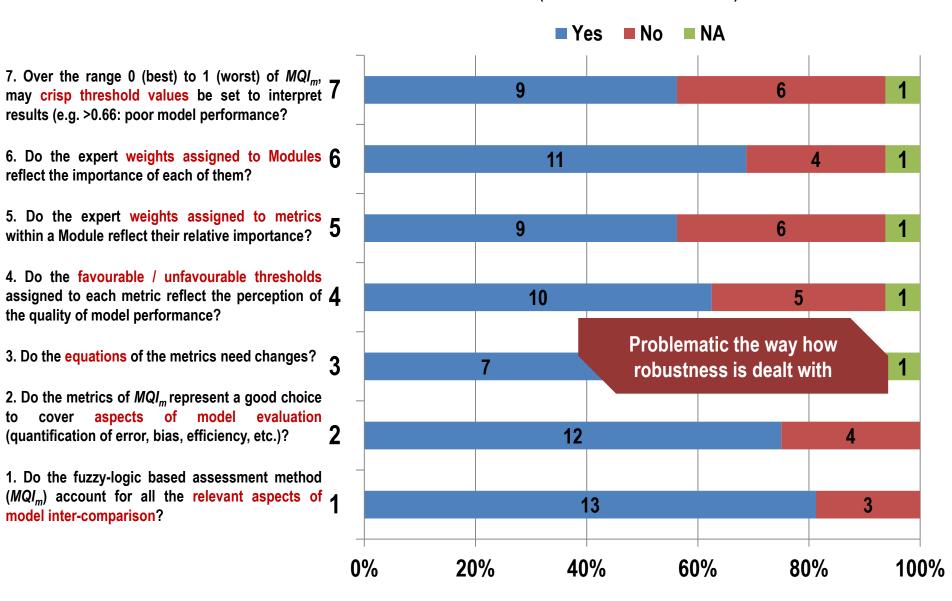
Index of robustness

Hindrances to overcome: thresholds and weights



MQI_m – Questionnaire

Questionnaires answered / commented: 16 (13 online + 3 offline) + 1 comment



Robustness of a model

A robustness measure would account for model performance stability over a wide range of conditions (single site versus multiple sites)

How the variability of model performance can be quantified with the variability of conditions?

Index of robustness

$$I_R = \frac{\sigma_{EF}}{\sigma_{SAM}}$$
 (0, best; +\infty, worst)

Modelling efficiency

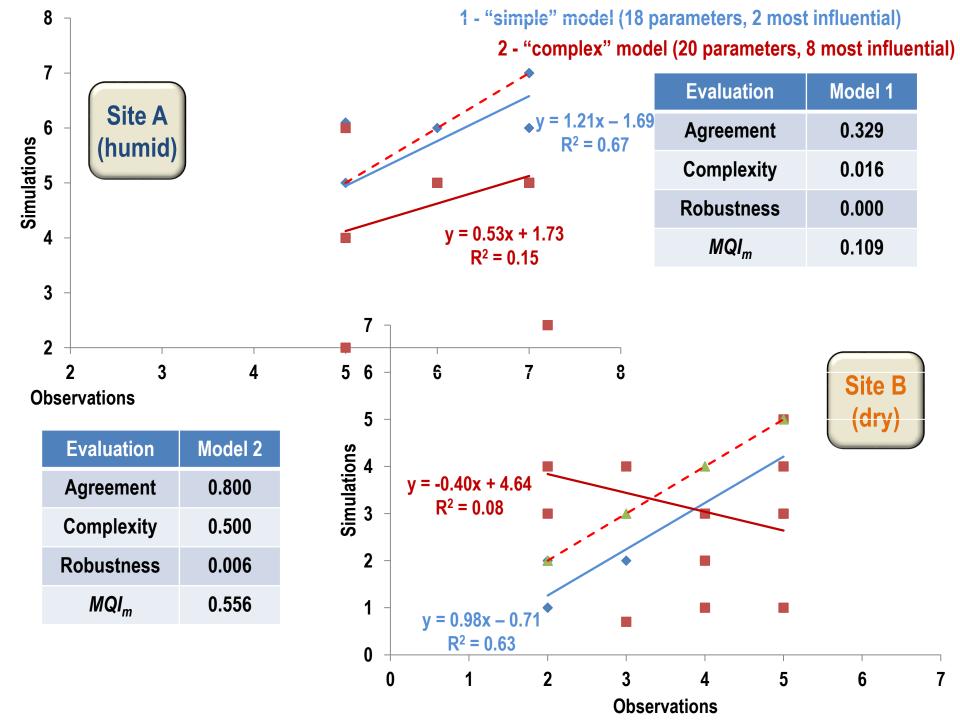
 $EF = 1 - \frac{\sum_{i=1}^{n} (P_i - O_i)^2}{\sum_{i=1}^{n} (O_i - \overline{O})^2} \text{ (-∞, worst; 1, best)} \qquad SAM = \frac{Rain - ET_0}{Rain + ET_0} \text{ (-1, +1)}$

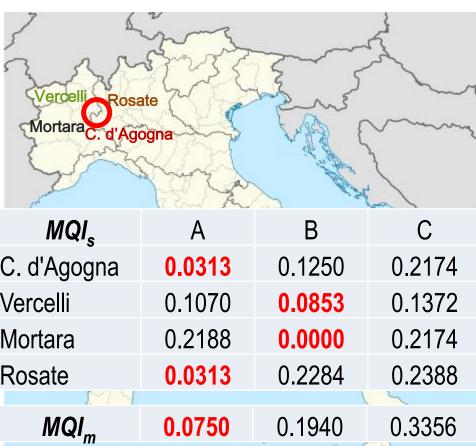
Synthetic Agro-Meteorological Indicator

$$SAM = \frac{Rain - ET_0}{Rain + ET_0} \quad (-1, +1)$$

From the questionnaires:

- Need to test the index on a variety of rainfall patterns (e.g. monsoonal areas)
- Whole year versus growing season, or winter and summer?
- Accounting for soil properties if water limited simulations are performed





Α

0.90

0.92

0.96

0.92

0.16

В

0.95

0.97

0.98

0.62

1.24

0.93

0.96

0.98

0.48

1.71

EF

 I_R

C. d'Agogna

Vercelli

Mortara

Rosate

Exemplary results

Above-ground rice biomass (kg DM m⁻²)

Three models: A (intermediate), B (simple), C (complex)

AIC	Α	В	С
C. d'Agogna	34	37	79
Vercelli	33	34	73
Mortara	26	28	67
Rosate	20	49	91

Complexity

Robustness

Towards an internationally-agreed protocol to evaluate C-N models: what does go forth?

- Review of settings
 - Selection of metrics,
 - Attribution of thresholds and weights

Extension to multiple outputs (three GHGs)



- Working group to be established
- Protocol to be elaborated





Thank you

for your attention!

