

Fuzzy-logic based multi-site crop model evaluation

Gianni BELLOCCHI

French National Institute for Agricultural Research, Clermont-Ferrand, France

Marco ACUTIS, University of Milan, Italy

Roberto FERRISE, University of Florence, Italy

Mike RIVINGTON, James Hutton Institute, Aberdeen, United Kingdom

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A Review of Methodologies to Evaluate Agroecosystem Simulation Models

F. MARTORANA and G. BELLOCCHI

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Review article

Validation of biophysical models: issues and methodologies. A review

Gianni BELLOCCHI^{1*,**}, Mike RIVINGTON², Marcello DONATELLI^{1***}, Koen



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Agricultural Systems

journal homepage: www.elsevier.com/locate/agr

An evaluation of the statistical methods for testing the performance
of crop models with observed data

J.M. Yang^a, J.Y. Yang^{b,*}, S. Liu^{b,c}, G. Hoogenboom^d

Assessment of the adequacy of

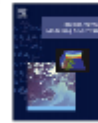


Environmental Modelling & Software 26 (2011) 328–336

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Technical assessment and evaluation of environmental models and software: Letter to the Editor

Alexandrov^{a,*}, D. Ames^b, G. Bellocchi^c, M. Bruen^d, N. Crout^e, M. Erechtkoukova^f, A. Hildebrandt^g,
J. Iqbal^h, J. J. Jonesⁱ, J. J. Jones^j, J. J. Jones^k, J. J. Jones^l, J. J. Jones^m, J. J. Jonesⁿ, J. J. Jones^o, J. J. Jones^p, J. J. Jones^q, J. J. Jones^r, J. J. Jones^s, J. J. Jones^t, J. J. Jones^u, J. J. Jones^v, J. J. Jones^w, J. J. Jones^x, J. J. Jones^y, J. J. Jones^z

Ecological Modelling 220 (2009) 1395–1410

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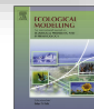
of the models WARM, CropSyst, and WOFOST for rice
Acutis^b, Gianni Bellocchi^c, Marcello Donatelli^{d,1}

Ecological Modelling 221 (2010) 960–964

Contents lists available at ScienceDirect

Ecological Modelling

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robustness based on the
explored conditions

Spanish Journal of Agricultural Research 2009 7(3), 680–686
ISSN: 1695-071-X

strategies for rice modelling
M. Boschetti³ and M. Acutis⁴

Expert System

Journal of
Applied Remote Sensing

Derivation of biophysical variables
Earth observation data: validation
statistical measures

Katja Richter
Clement Atzberger
Tobias B. Hank
Wolfram Mauser

An Indi

Elaboration of new metrics

Setting of thresholds

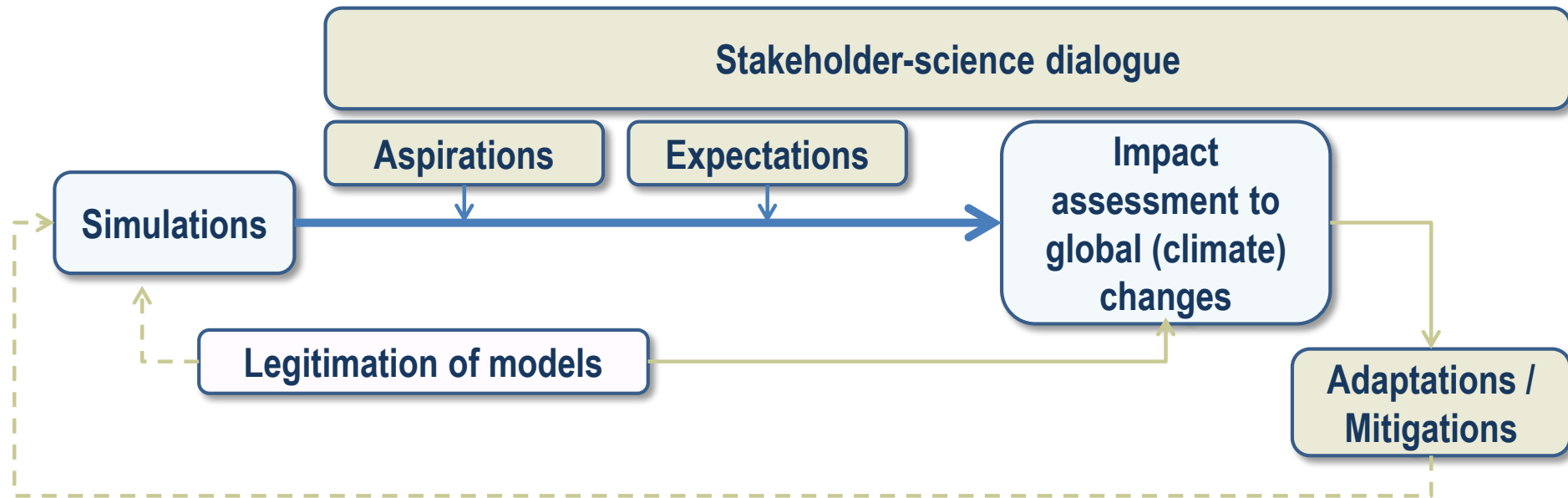
Meaning and limitations

Intercorrelation

Disaggregation

Aggregation

Deliberative process in model-based climate change studies

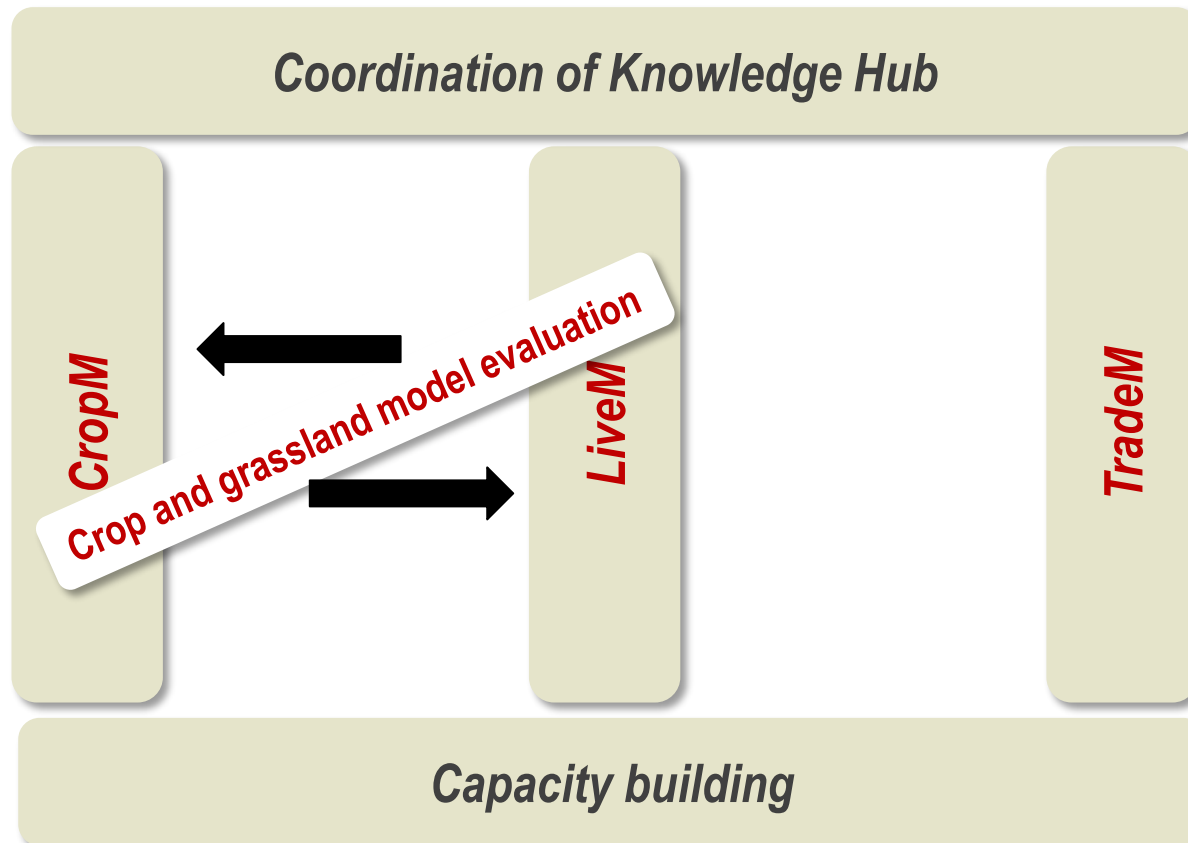


Bellocchi et al., 2006, Ital. J. Agrometeorol.

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

Bellocchi et al., 2015, Agron. Sustain. Dev.

MACSUR cross-cutting activities



CropM-LiveM

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

Some metrics

$$CRM = \frac{\sum_{i=1}^n O_i - \sum_{i=1}^n P_i}{\sum_{i=1}^n O_i}$$

$$MAE = \frac{\sum_{i=1}^n |P_i - O_i|}{n}$$

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (P_i - O_i)^2}{n}}$$

simple

absolute

squared

difference-based
metrics

non-parametric

$$MdAE = \text{median}_{i=1, \dots, n} |P_i - O_i|$$

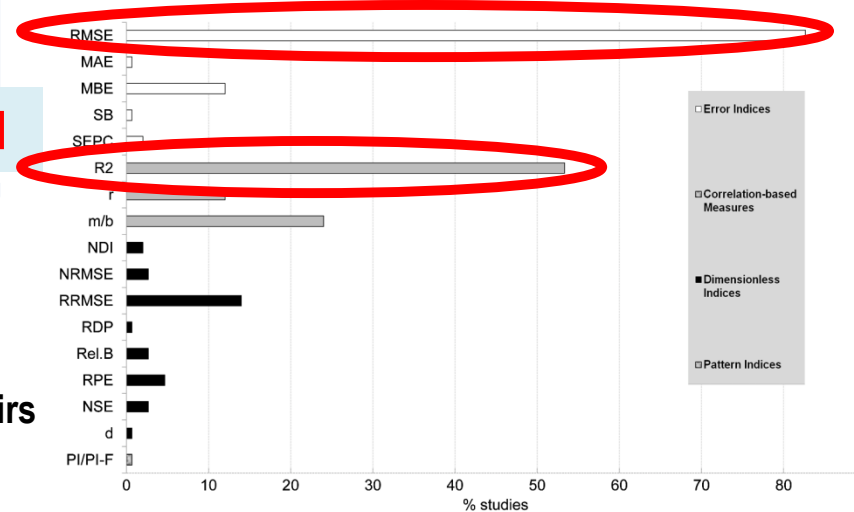
association-based metrics

$$r = \frac{\sum_{i=1}^n (P_i - \bar{P}) \cdot (O_i - \bar{O})}{\sqrt{\sum_{i=1}^n (P_i - \bar{P})^2 \cdot \sum_{i=1}^n (O_i - \bar{O})^2}} \quad r^2, \text{slope, intercept}$$

$$EF = 1 - \frac{\sum_{i=1}^n (P_i - O_i)^2}{\sum_{i=1}^n (O_i - \bar{O})^2}$$

$$d = 1 - \frac{\sum_{i=1}^n (P_i - O_i)^2}{\sum_{i=1}^n (|P_i - \bar{O}| + |O_i - \bar{O}|)^2}$$

combined



P = predicted; O = observed; i = ith O/P pair; n = number of O/P pairs

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

De Jager, 1994, Water SA

Key issues and factors

Key validation issues	Major factors to investigate				
	Modelling objective	Model inputs	Model outputs	Model structure	Modelling conditions
Validation purpose	X		X		X
Robustness of results			X		X
Interpretation of phenomena		X	X	X	
Model comparison				X	
Model predictions	X		X		X
Model complexity		X	X	X	
Data accuracy		X	X		
Time histories			X		

Fuzzy-logic based indicators

Model Quality Indicator (MQI_s)

MQI_s

membership function
 $S[x, a = 0; b = 1]$

membership function
 $S[x, a = \min(F, U); b = \max(F, U)]$

membership function
 $S[x, a = \min(F, U); b = \max(F, U)]$

Ratio of relevance parameters (R_p)
F Partial U
 $\geq 0.10 \leftrightarrow \leq 0.50$

A/C relative weight (w_k)
F Partial U
 $\geq 0.70 \leftrightarrow \leq 0.30$

Complexity
F Partial U
0 \leftrightarrow 1

Agreement
F Partial U
0 \leftrightarrow 1

Complexity

Agreement

F	F	0.00
F	U	0.75
U	F	0.25
U	U	1.00

0.00	F	F	F
0.20	F	F	U
0.60	F	U	F
0.80	F	U	U
0.20	U	F	F
0.40	U	F	U
0.80	U	U	F
1.00	U	U	U

Multi-site, Model Quality Indicator (MQI_m)

MQI_m

membership function
 $S[x; a = \min(F, U); b = \max(F, U)]$

membership function
 $S[x; a = 0; b = 1]$

Agreement

Complexity

Robustness

expert weight	Correlation coefficient (R) F Partial U $\geq 0.90 \leftrightarrow \leq 0.70$	Index of agreement (d) F Partial U $\geq 0.90 \leftrightarrow \leq 0.70$	Probability of equal means ($P(t)$) F Partial U $\geq 0.10 \leftrightarrow \leq 0.05$
0.00	F	F	F
0.20	F	F	U
0.60	F	U	F
0.80	F	U	U
0.20	U	F	F
0.40	U	F	U
0.80	U	U	F
1.00	U	U	U

	Ratio of relevance parameters (R_p) F Partial U $\geq 0.10 \leftrightarrow \leq 0.50$	AIC relative weight (w_R) F Partial U $\geq 0.70 \leftrightarrow \leq 0.30$
0.00	F	F
0.50	F	U
0.50	U	F
1.00	U	U

	Index of robustness (I_R) F Partial U 1 \leftrightarrow 10
0.00	F
1.00	U

	Complexity F Partial U 0 \leftrightarrow 1	Agreement F Partial U 0 \leftrightarrow 1	Robustness F Partial U 0 \leftrightarrow 1
0.00	F	F	F
0.25	F	F	U
0.50	F	U	F
0.75	F	U	U
0.25	U	F	F
0.50	U	F	U
0.75	U	U	F
1.00	U	U	U

membership function
 $S[x; a = \min(F, U); b = \max(F, U)]$

Synthetic indicators

*Aggregation rules:
fuzzy-logic based weighing system*

I. Agreement

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

II. Complexity

- Ratio of relevant parameters
- Parameters-agreement criterion

III. Stability (robustness)

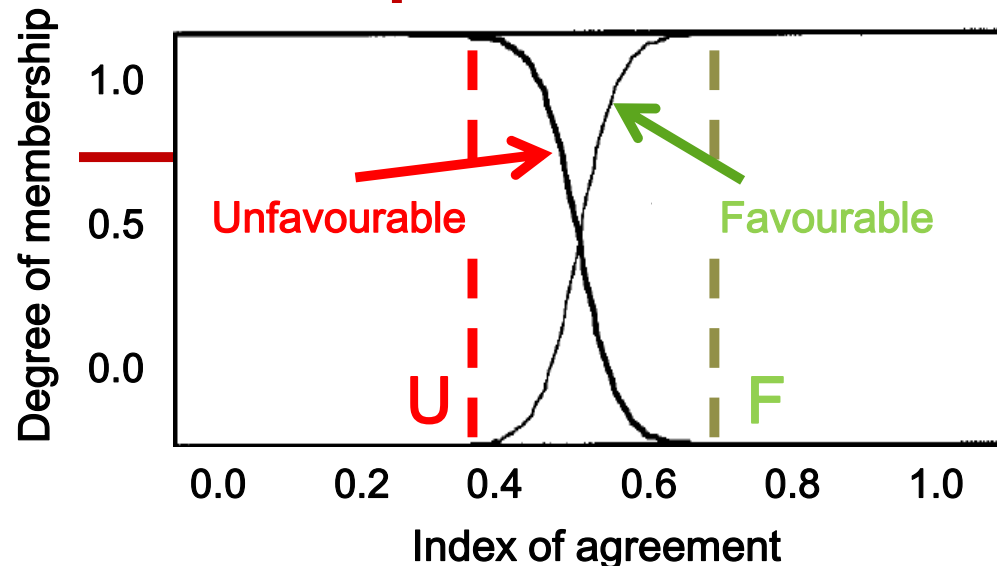
- Index of robustness

Non-dimensionality

Lower and upper bounding

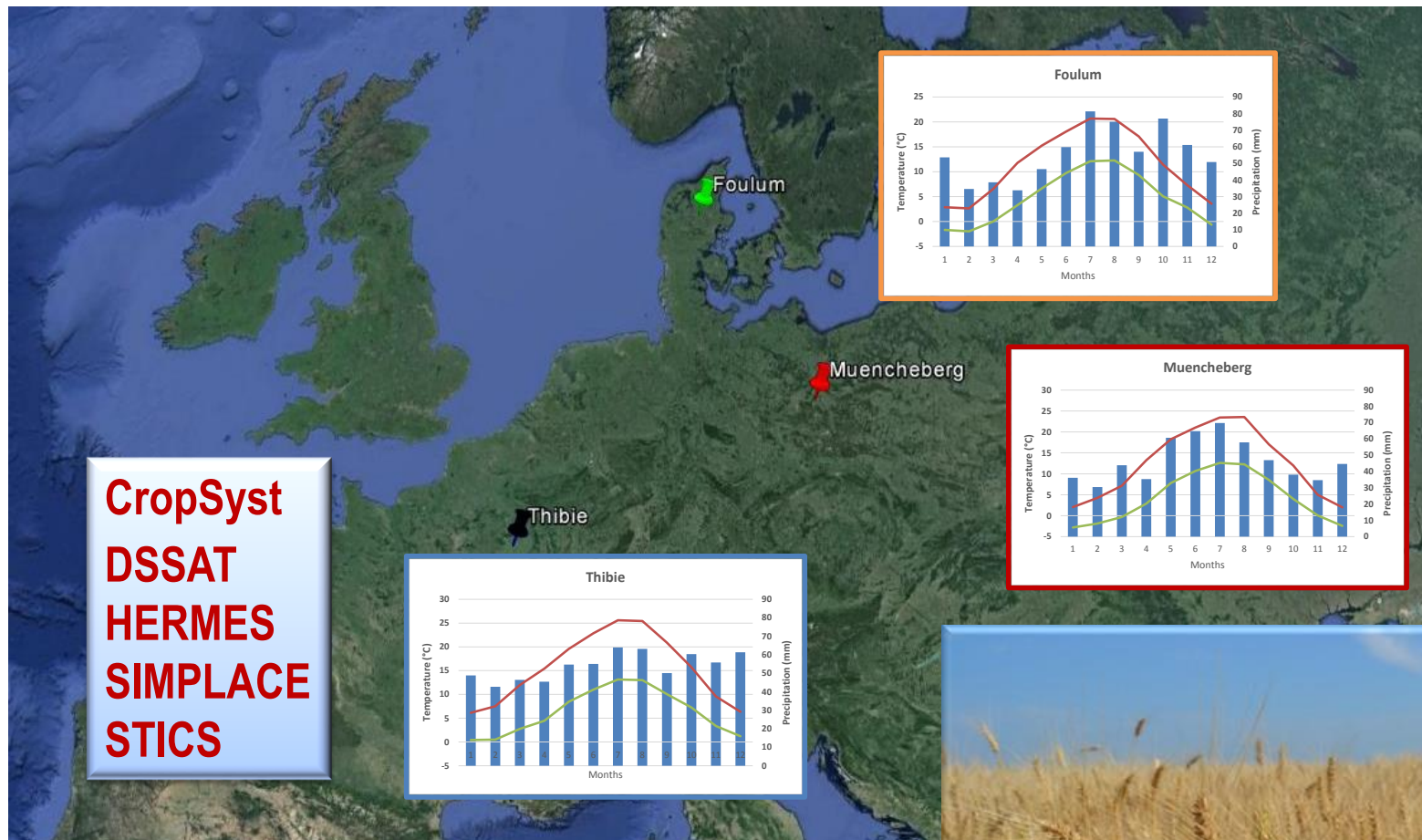
(best) 0 – 1 (worst)

Model Quality Indicator

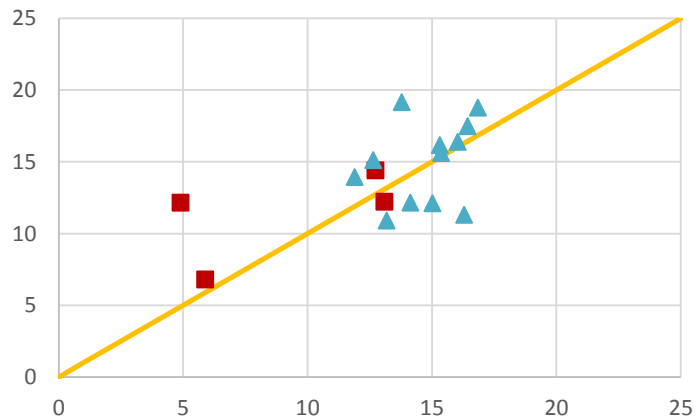


*Hindrances to overcome:
thresholds and weights*

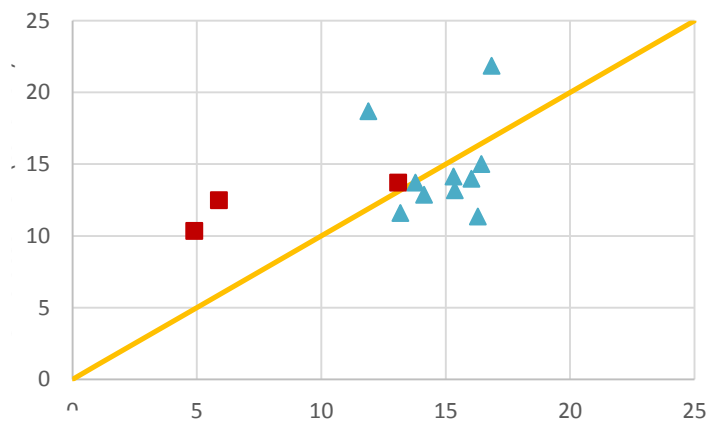
CropM wheat simulations: yield, above-ground biomass at maturity



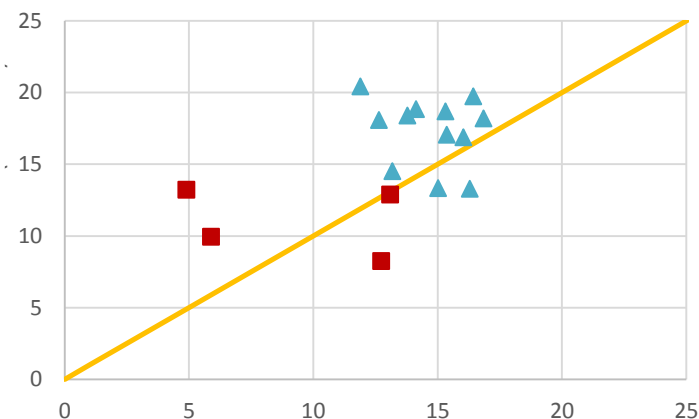
Model M1



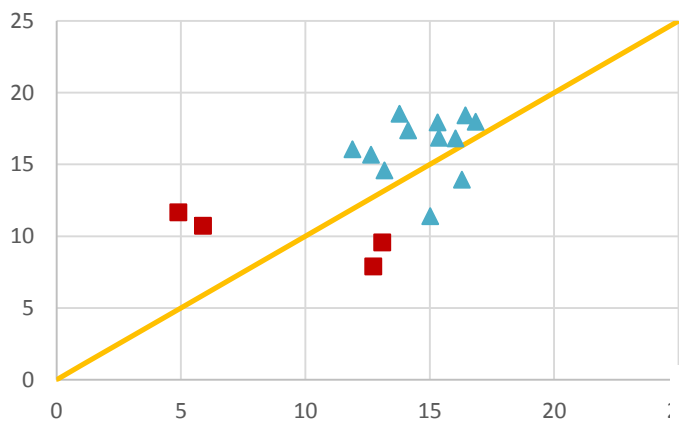
Model M2



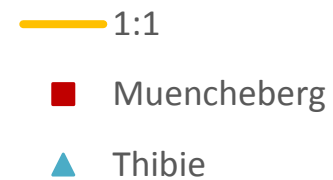
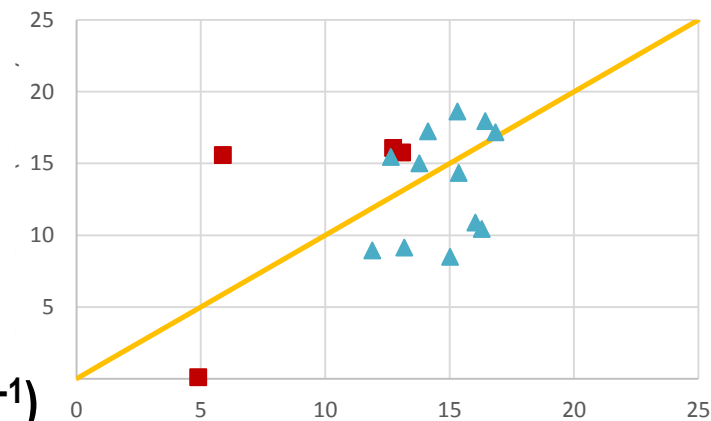
Model M3



Model M4

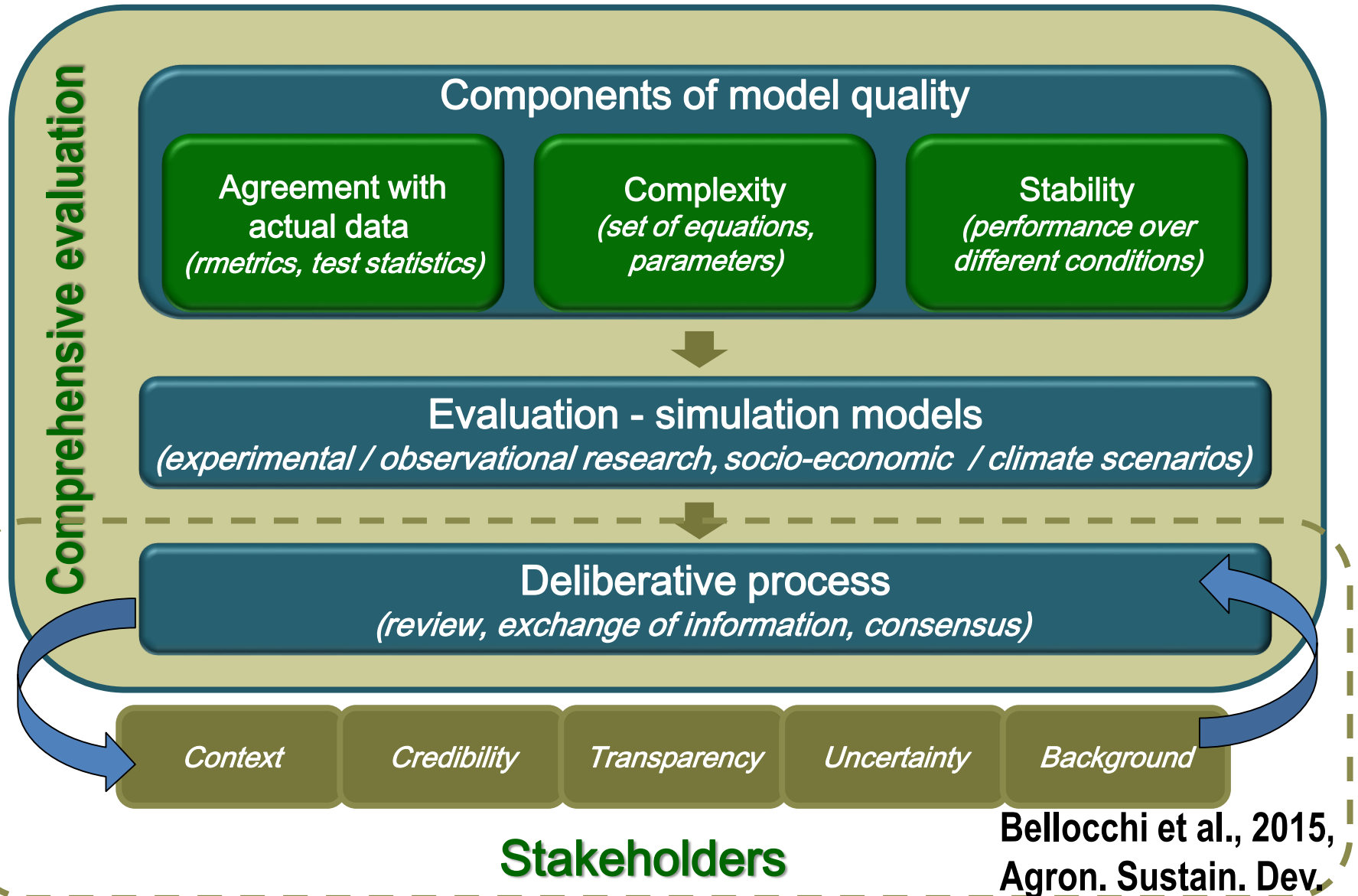


Model M5

Observed AGB (t ha⁻¹)Simulated AGB (t ha⁻¹)

Model	Aboveground biomass at maturity: performance metrics, modules and indicator					
	$\overline{P(t)}$	\bar{r}	\bar{d}	$\overline{R_p}$	$\overline{w_k}$	I_R
M1	0.23	0.46	0.64	0.32	1.99E-13	65.4
M2	0.20	0.46	0.60	0.28	2.66E-11	6.0
M3	0.01	-0.25	0.70	0.53	0.12	149.5
M4	0.08	-0.36	0.25	0.50	0.88	344.6
M5	0.08	0.49	0.60	0.37	1.34E-08	377.6
	Agreement			Complexity		Robustness
M1		0.8000		0.7975		1.0000
M2		0.8000		0.7975		0.6049
M3		1.0000		1.0000		1.0000
M4		0.8640		0.5000		1.0000
M5		0.8640		0.8944		1.0000
	MQI _m					
M1	0.8976					
M2	0.7471					
M3	1.0000					
M4	0.8428					
M5	0.9640					

Model evaluation / deliberative process

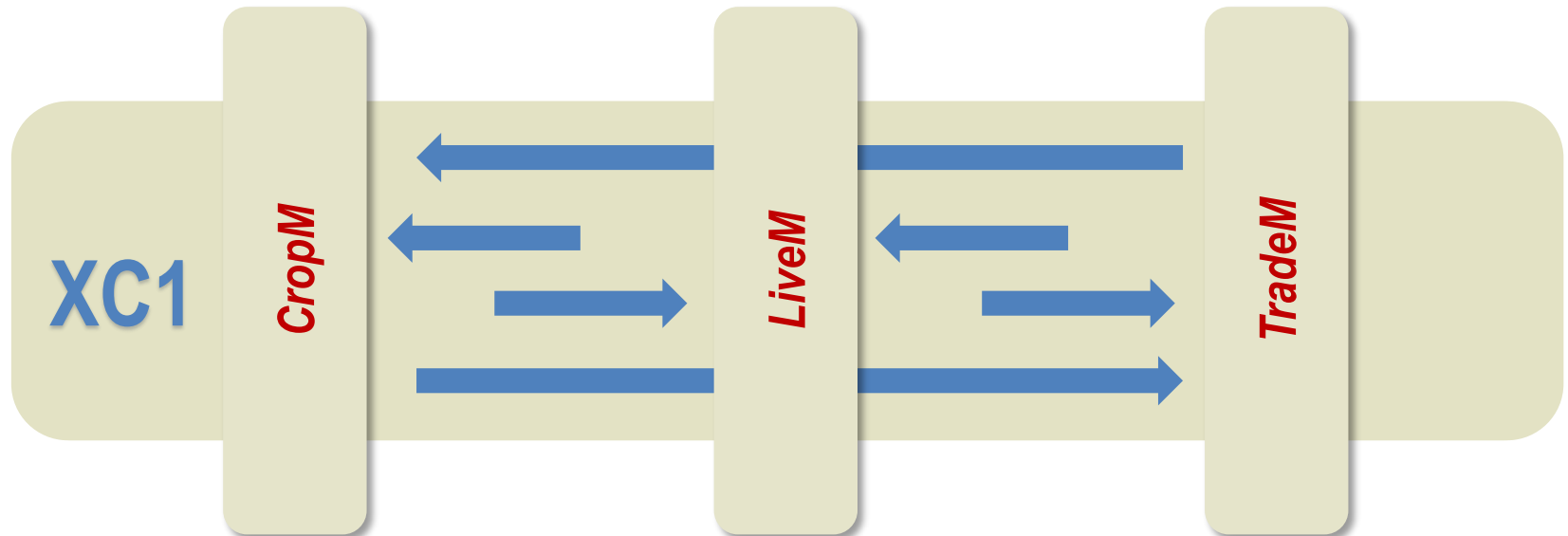


Towards a consolidated, internationally-agreed protocol to evaluate models: what does go forth?

❖ Review of settings

- ❖ Selection of metrics
- ❖ Attribution of thresholds and weights

❖ Extension to multiple outputs



Literature sources

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Acronym

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