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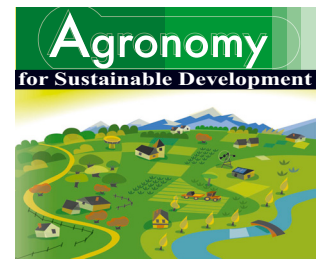
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## Research article

# MASC, a qualitative multi-attribute decision model for ex ante assessment of the sustainability of cropping systems

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**Abstract** – Realistic assessments of sustainability are often viewed as typical decision-making problems requiring multi-criteria decision-aid (MCDA) methods taking into account the conflicting objectives underlying the economic, social and environmental dimensions of sustainability, and the different sources of knowledge representing them. Some MCDA-based studies have resulted in the development of sustainable agricultural systems, but the new challenges facing agriculture and the increasing unpredictability of their driving forces highlight the need for faster ex ante (*‘Before-the-event’*) assessment frameworks. These frameworks should also (i) provide a more realistic assessment of sustainability, by integrating a wider range of informal knowledge, *via* the use of qualitative information; (ii) address alternative scales, such as cropping system level, improving granularity for the handling of sustainability issues and (iii) target a larger panel of decision-makers and contexts. We describe here the MASC model, which is at the center of a framework addressing these objectives. The MASC model has at its core a decision tree that breaks the sustainability assessment decisional problem down into simpler units as a function of sustainability dimensional structure (economic, social and environmental), generating a vector of 32 holistic ‘mixed’ (quantitative and qualitative) elementary criteria rating cropping systems. The assessment process involves the calculation of these criteria, their homogenization into qualitative information for input into the model and their aggregation throughout the decision tree based on *‘If-Then’* decision rules, entered by the user. We present the model and describe its first implementation for the evaluation of four cropping systems generated from expert knowledge, and discuss its relevance to the objectives cited above. The MASC model has several advantages over existing methods, due to its ability to handle qualitative information, its transparency, flexibility and feasibility.

**cropping system / sustainability assessment / ex ante / MCDA / decision rules / qualitative multi-attribute decision models / DEXi**

## 1. INTRODUCTION

The agricultural sector is currently at the confluence of an increasing number of new challenges such as market globalization, food crisis, bio-fuels, environmental concerns, and changes in legislation at the global and local scales that are threatening the future of agricultural systems in many areas of the world. Almost two decades ago the foreseeable uncertainties and threats conditioning the future of agricultural systems resulted in the development of the *agricultural sustainability* concept, which has been defined in many ways (Hansen,

1996; Christen, 1998), all highlighting its holistic and multi-dimensional nature and encompassing economic, social and environmental objectives.

Ever since the emergence of this concept, it has been claimed that the assessment of multidimensional sustainability is required for the implementation of new forms of sustainable agriculture (Neher, 1992; Schaller, 1993; Vereijken, 1997; den Biggelaar and Suvedi, 2000; Pacini et al., 2003; Gafsi et al., 2006). The first operational studies concluded that, for such assessments to be realistic, a ‘methodological leap’ was needed (i) to integrate knowledge of different types relating to economic, social and environmental objectives and (ii) to handle the conflicting aspects of these objectives (Munda et al.,

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1994; Dent et al., 1995). One alternative involved considering sustainability assessment as a typical decision-making problem that could be handled by multi-criteria decision-aid (MCDA) methods (Munda et al., 1994; Munda, 2005). As a result, over the last ten years, a number of MCDA-based frameworks have been developed to facilitate the implementation of sustainable forms of agriculture (Rossing et al., 1997; Zander and Kächele, 1999; Loyce et al., 2002; Dogliotti et al., 2004, 2005).

However, the current pace at which new challenges are appearing and shaping agricultural sustainability and the increasing unpredictability of the driving forces behind them make it necessary to find alternative ways to address the questions of the representation and assessment of sustainability. First, these fluctuations make it increasingly necessary to develop fast ex ante (i.e., ‘*Before-the-event*’) sustainability evaluation frameworks that can represent and assess a large body of options *in silico* and rapidly identify innovative, alternative systems without the need for in-field assessments of all the possible options (European Commission, 2005; Van Ittersum et al., 2007; Meynard, 2008).

Second, there is a need to improve the ‘granularity’ of the spatio-temporal scales generally used in sustainability assessment studies, making it possible to target specific sustainability issues more efficiently. The smallest spatiotemporal unit addressed in sustainability assessment studies is generally the farming system (e.g., Häni et al., 2003; Dogliotti et al., 2005; Meyer-Aurich, 2005; Meul et al., 2008), whereas much of the impact of the environment and the impact of certain socio-economic factors on farms are deeply rooted within smaller units, such as that of the cropping system — the level at which they are determined. These units, defined by Sebillotte (1990) as ‘*a set of management procedures applied to a given, uniformly treated area, which may be a field, part of a field or a group of fields*’ are rarely considered in assessments of sustainability (Rosnoble et al., 2006). Furthermore, in the few cases in which this scale has been specifically addressed (e.g., Mazzetto and Bonera, 2003), the number of sustainability criteria considered is strongly limited with respect to the holistic nature of the concept of sustainability. A more detailed assessment of cropping systems may make it possible to target and improve each of the systems making up the farming system, in a specific manner.

Third, many of the available MCDA-based frameworks are restrictive in terms of the type of information required and the types of decision-maker targeted. These restrictions on the type of information result from diverse contexts that cannot be addressed quantitatively being overlooked, thus limiting the range of innovative options assessed (Sadok et al., 2008). The restrictions on the type of decision-maker arise from the underlying decisional problem being addressed in a manner appropriate for only a small subset of decision-makers in many of these studies. Quantitative preference models, with formalisms that are ‘opaque’ or inaccessible to non-expert decision-makers, are often used. Dent et al. (1995) suggested that an assessment method is likely to be successful only if it uses the language of the decision-maker and represents all

aspects of his knowledge, including empirical, non-formal aspects.

In a comparative review of the main MCDA families focusing on relevance for sustainability assessment, Sadok et al. (2008) suggested that ‘mixed’ or ‘non classical’ MCDA methods were particularly relevant for handling the above constraints, while being able to address more efficiently the multi-dimensional constraints of sustainability assessment (i.e., incomparability, non compensation and incommensurability of dimensions). This is particularly true for decision rule-based methods, which allow the use of qualitative input information and qualitative reasoning in the decision modeling process. Programs using these methods — often in the form of decision support systems (DSS) — are increasingly designed for use by more diverse decision-makers, with the provision of transparent communication interfaces (Matthies et al., 2007). Some of these approaches have been used successfully to assess specific problems in agricultural systems, such as the impact of these systems on soil quality (Bohanec et al., 2007) or the ecological/economic impact of genetically modified crops (Bohanec et al., 2008). However, to our knowledge, the use of such approaches to assess explicitly and simultaneously the social, economic and environmental dimensions of the sustainability of a given agricultural system — in this case cropping systems — has never been reported.

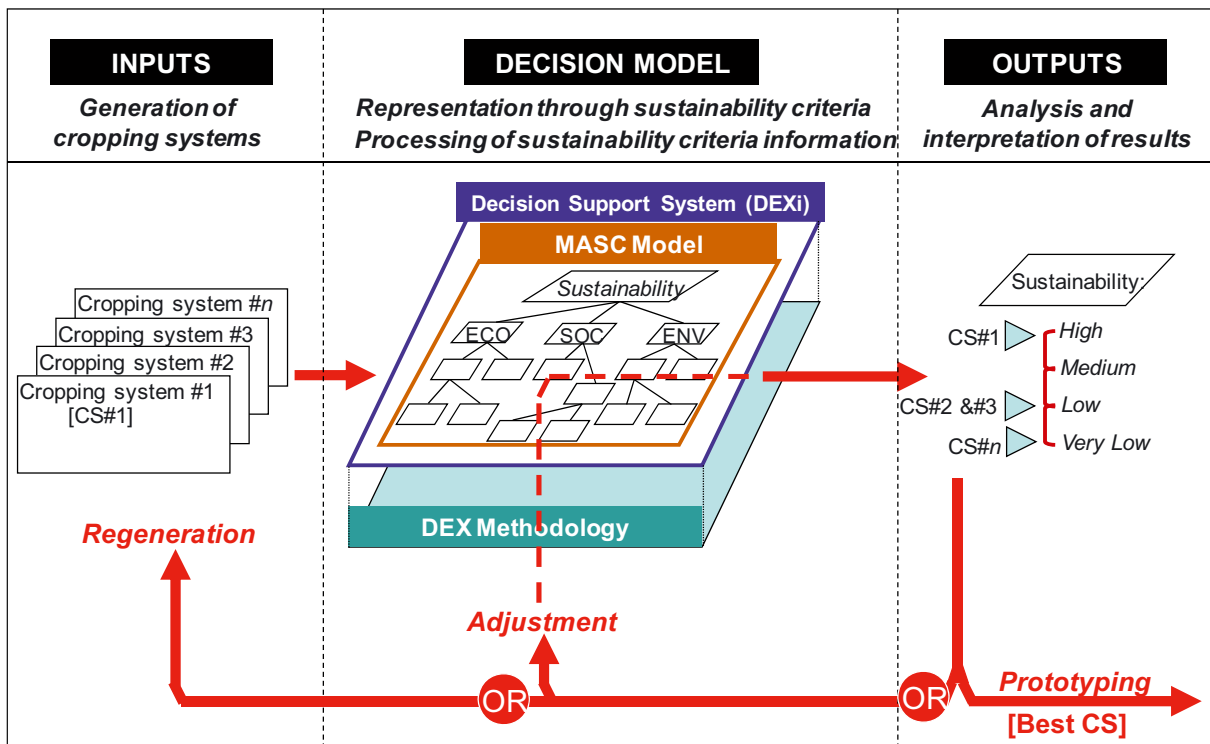
We describe here the *MASC* (for *multi-attribute assessment of the sustainability of cropping systems*) decision model, which was developed within a decision support system (DEXi), as a key element of a framework combining the ex ante generation and assessment of cropping systems. We present and discuss the decision model and its first implementation for the assessment of four cropping systems generated from expert knowledge.

## 2. METHODS

### 2.1. The assessment framework

The overall assessment framework consisted of a three-step approach (Fig. 1):

1. The generation of cropping systems, as inputs for the decision model. This step can be achieved with special computing tools (Dogliotti et al., 2003; Loyce et al., 2002; Bergez et al., 2008), based on expert knowledge (Lançon et al. 2008), or both.
2. Representation of each of the generated cropping systems as a vector of sustainability criteria expressed in qualitative terms and the processing of this vector by *MASC* to obtain a synthetic qualitative assessment of the overall (economic, social and environmental) sustainability of the cropping system.
3. Analysis and interpretation of the assessment results for the selection of alternative systems to be tested in field trials. If the decision-makers are not satisfied with the results, the framework offers the possibility of either ‘adjusting’ certain decision rules or, if necessary, testing a new set of improved cropping systems (Fig. 1).



**Figure 1.** Structure of the framework generating and assessing ex ante the sustainability of cropping systems. After generation (*in silico* through models, from expert knowledge or both), the cropping systems (inputs) are represented as a vector of values for sustainability criteria, which are processed by the MASC decision model, implemented within a decision support system (DEXi). Depending on the evaluation results (outputs), a cropping system may be (i) selected for prototyping in the field (ii) re-submitted to the decision model with ‘adjusted’ choices in the evaluation process or (iii) discarded/improved, possibly requiring a new generation process.

## 2.2. Methodological basis of MASC

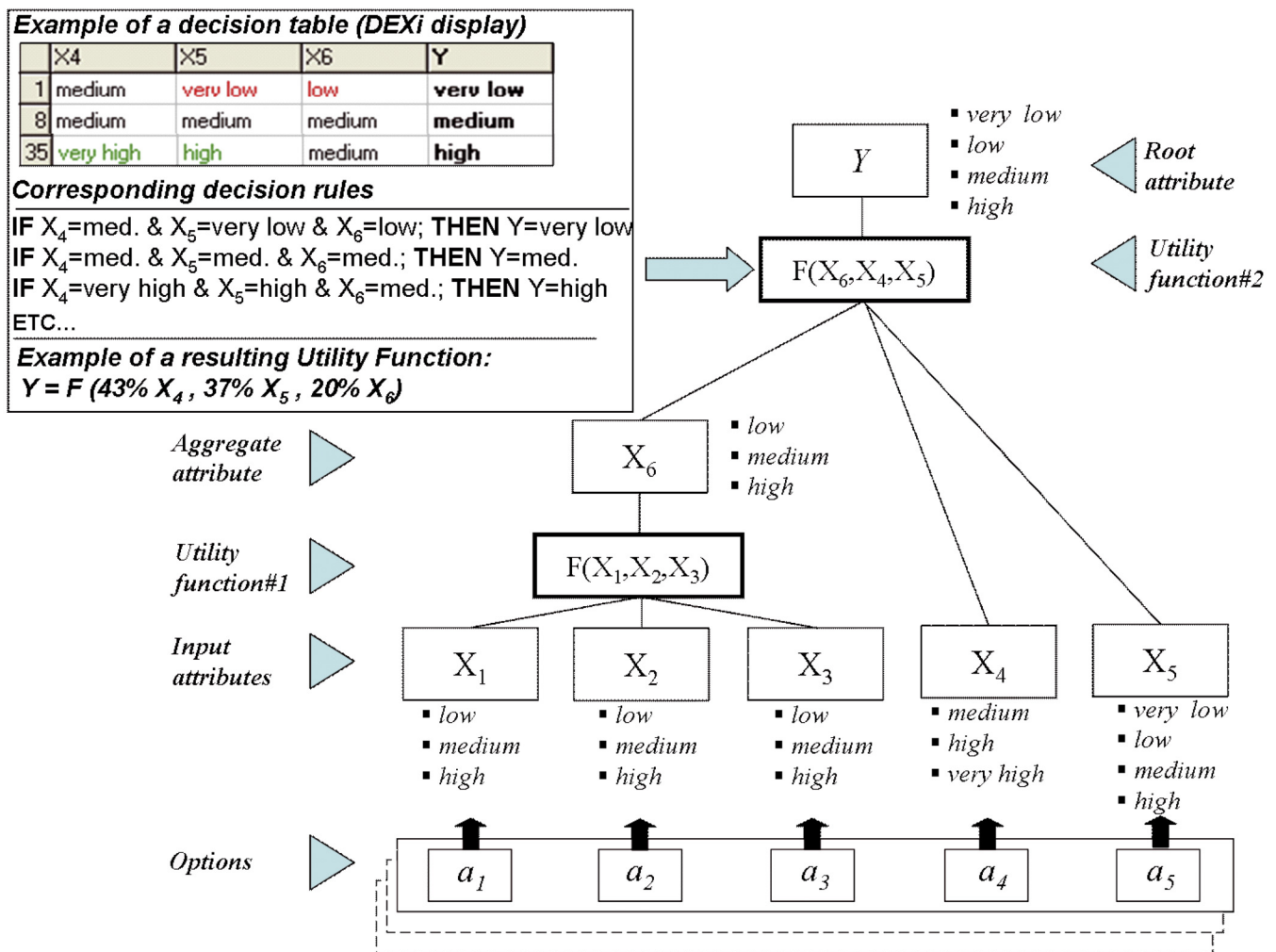
MASC is based on DEX methodology, and is implemented within a decision support system (DSS) called DEXi (Bohanec, 2003). It combines a hierarchical multi-attribute decision model and an expert system shell. Hierarchical multi-attribute decision models are used in many classical MCDA approaches, such as the AHP (Analytic Hierarchy Process) or other MAUT/MAVT (Multi-Attribute Utility/Value Theory) methods (Saaty, 1980; Clemen, 1996; Figueira et al., 2005). They make it possible to break a decisional problem (a *root attribute*) down into smaller, less complex subproblems represented by variables or *attributes* ( $X_i$ , Fig. 2). These attributes are organized hierarchically so that those at higher levels of the hierarchy depend on those at lower levels. Attributes are aggregated through *utility* (or aggregation) *functions*  $F$ , which determine the dependence of a given attribute (called *aggregate attribute*) on its immediate descendants in the hierarchy (see example in Fig. 2). *Options* or *alternatives* to be evaluated are represented by a vector of values ( $a_i$ ) of attributes, called *input attributes*.

The aggregation procedure followed by DEX methodology is based on the hierarchical multi-attribute representation of decision problems used in many classical MAUT/MAVT models, but differs from this representation in a number of ways. MAUT/MAVT methods generally use (i) aggregation proce-

dures based on multiplicative or additive formalisms that do not realistically capture the multidimensional nature of sustainability assessment and (ii) only quantitative information, hampering the use of empirical, non quantifiable knowledge that may be relevant for sustainability assessment (Munda et al., 1994; Munda, 2005). By contrast, DEX methodology is based on (i) qualitative (symbolic) attributes and (ii) utility functions based on ‘*If-Then*’ decision rules for the aggregation of these attributes. Such features make this approach highly suitable for realistic assessments of the sustainability of agricultural systems (Dent et al., 1995).

A DEX-based evaluation process follows the following pathway (Fig. 2):

- (i) A vector of input attributes is identified and the decision problem is represented through a hierarchy of attributes;
- (ii) Each attribute of the tree is scored on a qualitative scale (e.g., ‘*Low*’, ‘*Medium*’, ‘*High*’);
- (iii) Decision rule-based utility functions are established for each aggregate attribute. In DEXi, a table is made available for users to enter qualitative values for aggregate attributes (e.g.,  $Y$ , Fig. 2), as a function of the qualitative values of the attributes to be aggregated (e.g.,  $X_4$ ,  $X_5$  and  $X_6$ ), thereby defining decision rules (e.g., IF  $X_4 = \text{‘medium’}$  &  $X_5 = \text{‘very low’}$  &  $X_6 = \text{‘low’}$ ; THEN  $Y = \text{‘low’}$ ). Once the table is filled with these rules, the



**Figure 2.** Typical structure and functioning of a decision rule-based, qualitative hierarchical multi-attribute model (adapted from Bohanec et al., 2000). The inset in the top left corner provides an example of the way in which a utility function is developed in the decision support system DEXi (see main text for details).

software provides users with a utility function representing a compilation of their choices expressed in terms of relative weightings (i.e., aggregation rules);

- (iv) For the evaluation of an option, users enter the corresponding qualitative values of the input attributes and a qualitative result depending on the structure of the decision tree and the predefined utility functions is displayed (value of  $Y$ ).

In the DEXi decision support system, structures such as those shown in Figure 2 are dynamic, in that changing the qualitative value of a given attribute and/or modifying a few decision rules can have an immediate effect on the overall assessment. This approach makes it possible to carry out ‘*what-if*’ and ‘*selective explanation*’ and sensitivity analyses. In MASC, input attributes represent sustainability criteria and provide a qualitative rating of each cropping system (option). The aggregate attributes are represented by the various levels of sustainability assessment subproblems addressed by the decisional model. In

its current version, the model structure, sustainability criteria and attribute scale structures are predefined. Aggregation rules are left open for definition by the user in  $\approx 70\%$  of cases (with suggested thresholds), for adaptation of the model to the climatic and agricultural context, with the remaining 30% predetermined on the basis of expert knowledge (Sect. 2.4.2). The qualitative rating of sustainability criteria is left to the decision-maker (with suggested guidelines, Sect. 2.4.1). The steps leading to the development of MASC are outlined in the section below.

### 2.3. Workgroup organization and methodology

The development of MASC and the first examples of its implementation result from the collaboration of a workgroup that functioned over two years according to the specific organizational and methodological procedures presented below.

### 2.3.1. Model development

The development of DEX/DEXi-based models requires iteration of the following four steps: (i) attribute identification; (ii) attribute structuring; (iii) the definition of qualitative measurement scales for each attribute and (iv) the definition of aggregation rules. Typically, such models require collaboration between experts in the considered fields, who suggest attributes and aggregation rules, and decision analysts, who carry out the process and define a hierarchical structure for the attributes (Bohanec et al., 2008).

In our study, these four steps were carried out by eight agronomists. Each contributed his or her own expertise and, when necessary, exchanges took place with a pool of about 25 experts in research and development who took part in a previous program (Reau and Landé, 2006; Reau et al., 2006). Each of the eight agronomists also acted as a decision analyst, allowing the group to explore, compare and contrast different possible structures for the initial decision model and to identify the most consensual and relevant model. Input attributes were selected/developed by the workgroup on the basis of two main considerations:

(1) ‘*Thematic relevance*’, corresponding to the ability of the attributes to address realistically all the relevant issues underlying the assessment of sustainability at the scale of the cropping system. This goal was achieved by carrying out a specific literature review for each of the issues considered, to determine whether an indicator or methodology for its estimation had already been identified.

(2) ‘*Operational relevance*’, corresponding to the feasibility of implementing the chosen set of criteria in a multi-criterion evaluation approach without biasing the assessment results. This goal was achieved by following the recommendations of Keeney and Raiffa (1976), Maystre et al. (1994) and Baker et al. (2002). We evaluated the extent to which the selected criteria were: (i) exhaustive and complete (i.e. including all goals), (ii) operational and meaningful in terms of the decision-maker’s understanding of the implications of the assessment, (iii) able to discriminate between cropping systems, (iv) non redundant (as far as possible) and (v) realistically few enough in number to keep problem dimensions manageable.

### 2.3.2. First examples of implementation

The four cropping systems used to test the MASC model were initially developed based on the knowledge of 25 experts. One cropping system was designed to be representative of the cropping systems currently prevailing in Picardy/northern France (cropping system #1), whereas the other three represented possible alternatives (cropping systems #2, #3 and #4; Reau and Landé, 2006; Reau et al., 2006). The first MASC-based assessment process for these systems was carried out by the group of eight agronomists, acting as decision analysts.

## 2.4. Features of the MASC model

The overall structure of the MASC model is shown in Figure 3. It includes two interdependent components: (i) an input information processing unit (Fig. 3A) and (ii) a decision tree (Fig. 3B). These two components define a decision rule-based, qualitative multi-attribute model.

### 2.4.1. Input information processing

The input information (i.e., sustainability criteria) of the model is structured into a vector of 32 input attributes (Tab. I, Fig. 3). This vector was developed according to the working procedures described in Section 2.3. The vector had a heterogeneous initial informational structure, combining qualitative and quantitative information. The input information processing phase (i) calculates/estimates the values of the input attributes based on the corresponding methodology (Tab. I) and (ii) homogenizes the values by converting them all into qualitative (linguistic) variables, for processing by the decision tree (Fig. 3A).

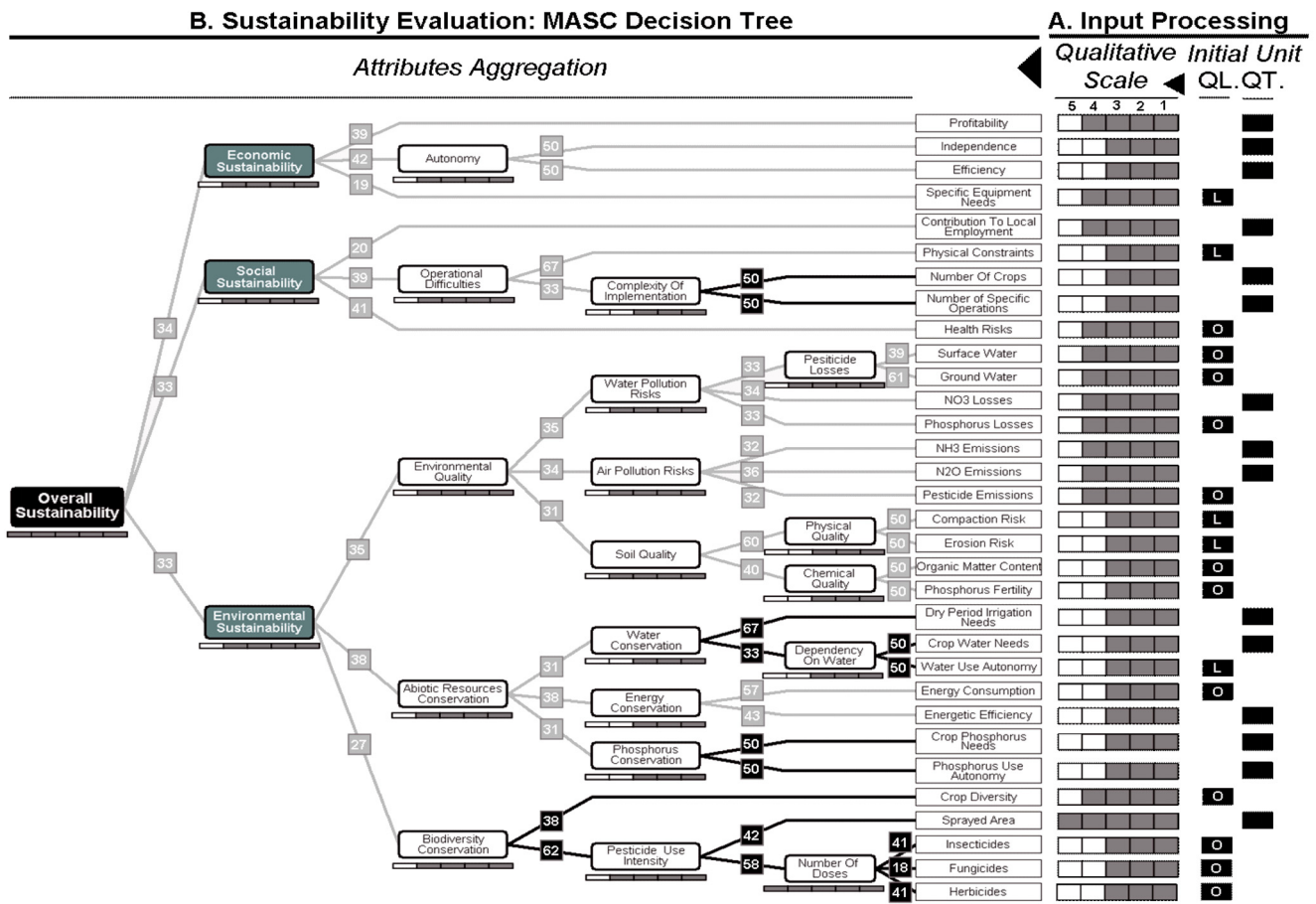
#### 2.4.1.1. Initial calculation/estimation of the input attributes

Table I summarizes the methods used to calculate the input attributes. All the calculations/estimations are carried out for each year of the rotation and averaged, except for one approach (I-Phy indicators, see below for details). During this phase, we distinguished between two main groups of attributes, based on the initial type of information (Fig. 3A).

The first group consisted of 15 input attributes initially estimated *quantitatively*, based on formalisms and models adapted from previous studies (e.g., economic attributes, *Contribution to Local Employment*, *NH<sub>3</sub> Emissions*), or developed directly from expert knowledge (e.g., *Phosphorus Use Autonomy*, *Sprayed Area*, Tab. I).

Where necessary, formalisms and models from published studies were adapted to the scale of the cropping system. This was the case, for example, for economic attributes, such as *Semi-Net Margin*, which was adapted from the work of Dogliotti et al. (2004) for the estimation of cropping system profitability. Costs due to the storage, maintenance and repair of machinery, and infrastructure costs for buildings, roads and fences — which are not relevant at the cropping system scale — were not included in calculations of operating costs. Empirical quantitative models were developed, based on the expertise of the workgroup and the pool of consulted experts, if no satisfactory approach had been described in previous studies. This was the case for the attribute estimating the level of *Phosphorus Use Autonomy* of the cropping systems, as many studies have shown that phosphorus resources are not renewable and gradually decrease (Steen, 1998; Stewart et al., 2005).

The second group consisted of 17 input attributes initially estimated *qualitatively*, through two different approaches. The first approach involved the use of indicators readily giving



**Figure 3.** Sustainability criteria information processing and aggregation by the MASC decision tree, allowing ex ante assessment of the sustainability of cropping systems. A. After calculation/estimation (see Tab. I for details), initial units of the input attributes — either quantitative (QT, black rectangles) or qualitative (QL, black rectangles with L for ‘linguistic’ and O for ‘ordinal’) — are homogenized into linguistic values (e.g., ‘Low’, ‘Medium’, ‘High’) on 3- to 5-point scales (represented by the number of gray rectangles located to the right of the input attributes). B. The input attributes are then aggregated through levels of aggregate attributes (rounded rectangles), where the black and the three gray rectangles represent the overall and one-dimensional (economic, social or environmental) evaluation results, respectively. The aggregation rules are of two types: (i) totally free (thick gray lines) when aggregation is dependent on the specific context of the cropping system and on the knowledge and views of the decision-maker; and (ii) predetermined (thick black lines) when aggregation is independent of the context and depends instead on factual/expert knowledge (see details in the text). Numerical values in the decision tree displayed in gray and black boxes represent weightings corresponding to types (i) and (ii), respectively. They represent the aggregation rules used for the evaluation of four examples of cropping systems, the results of which are reported in Table II and Figure 4.

qualitative-ordinal scores (‘O’ rectangles in Fig. 3A), including certain indicators from the I-Phy/INDIGO method (Bockstaller et al., 1997; Bockstaller and Girardin, 2006; e.g., *Surface Water* and *Ground Water* pesticide losses, soil *Organic Matter Content*, *Energy Consumption*), and those stemming from the treatment frequency index approach (Champeaux, 2006; e.g., *Health Risks*, number of doses for *Insecticides/Fungicides/Herbicides*). In this group, the inclusion of values for I-Phy indicators at the rotation scale was not directly based on averaging yearly values (as for all the other attributes), and the specific development of a ‘satellite’ decision tree (not shown) was required for calculation.

A second approach involves the development of bibliographic or expertise-based guidelines for the direct

qualitative-linguistic estimation (e.g., ‘Low’, ‘Medium’, ‘High’) of attributes addressing issues for which no satisfactory quantitative or qualitative model could be identified in previous studies (‘L’ rectangles in Fig. 3A). This was the case, for example, for the *Water Use Autonomy* attribute, which estimates the extent to which water requirements depend on the main type of water supply to the cropping system.

**2.4.1.2. Qualitative transformation: scale and value choice**

Once calculated or estimated, the initial values of all input attributes are transformed into a qualitative-linguistic

**Table I.** Summary of the different knowledge sources used as input attributes in MASC. <sup>1</sup> Input attribute name (attribute order is the same as that reported in Fig. 3). <sup>2</sup> Approaches used to calculate/estimate the input attributes. Quantitative formalisms are either described textually or mathematically when simple (e.g. water resource attributes and economic attributes, respectively) or referred to when too complex to fit in the table (e.g., some Indigo indicators, see corresponding references). Similarly, qualitative approaches are either described in text (when not based on specific formalisms) or referred to when based on published indicators. Calculations/estimations are made for each year of the rotation then averaged, except for I-Phy indicators (see text for details). Units are indicated when necessary. Units of qualitative attributes can be expressed on the basis of qualitative-ordinal scales (Indigo indicators) or qualitative-linguistic scales. <sup>3</sup> References used. a: Dogliotti et al. (2004); b: Vilain et al. (2008); c: MSA (2006); d: Bockstaller et al. (1997), Bockstaller and Girardin (2006); e: Spoor et al. (2003), van de Zande (1991); f: Champeaux (2006); X: expertise. When necessary, imported formalisms were adapted to the scale of the cropping system (see text for details).

Input Attribute <sup>1</sup>	Description / Calculation Method <sup>2</sup>		Reference <sup>3</sup>
<i>Profitability</i>	SNM = GP - OC	SNM: Semi-Net Margin (€ ha <sup>-1</sup> ) GP: Gross Product (€ ha <sup>-1</sup> ) OC: Operational Costs (€ ha <sup>-1</sup> )	a
<i>Independence</i>	EI = (1-DS/SNM) × 100	EI: Economic Independence (%) DS: Direct Subsidies (€ ha <sup>-1</sup> )	b
<i>Efficiency</i>	EE = (1-OC/GP) × 100	EE: Economic Efficiency (%)	b
<i>Specific Equipment Needs</i>	Qualitative estimation		X
<i>Contribution To Local Employment</i>	NTW = EMWU / S	NTW: Number of temporary workers ha <sup>-1</sup> EMWU: Number of external man work units	b
<i>Physical Constraints</i>	Qualitative estimation based on: vibrations frequency, mouvements repetitiveness, frequency of heavy loads manipulation, allergies, noise levels		c
<i>Number Of Crops</i>	Indicator of implementation complexity		X
<i>Number Of Specific Operations</i>	Indicator of implementation complexity		f
<i>Health Risks</i>	TFI <sub>Haz</sub> : Treatment Frequency Index calculated for hazardous products		X
<i>Surface Water</i>	I-Phy <sub>SW</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>Ground Water</i>	I-Phy <sub>GW</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>NO<sub>3</sub> Losses</i>	I-NO <sub>3</sub>	Indigo Method (kg NO <sub>3</sub> -N ha <sup>-1</sup> )	d
<i>Phosphorus Losses</i>	I <sub>P</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>NH<sub>3</sub> Emissions</i>	I <sub>NH<sub>3</sub></sub>	Indigo Method (kg NH <sub>3</sub> -N ha <sup>-1</sup> )	d
<i>N<sub>2</sub>O Emissions</i>	I <sub>N<sub>2</sub>O</sub>	Indigo Method (kg N <sub>2</sub> O-N ha <sup>-1</sup> )	d
<i>Pesticide Emissions</i>	I-Phy <sub>air</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>Compaction Risk</i>	Qualitative estimation based on texture class, mechanization level, soil wetness, soil work type		e
<i>Erosion Risk</i>	Qualitative estimation based on soil water status, crusting sensitivity, cover type, soil work type		X
<i>Organic Matter Content</i>	I <sub>MO</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>Phosphorus Fertility</i>	I <sub>Psoil</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>Dry Period Irrigation Needs</i>	Indicator of water resource conservation capacity (mm)		X
<i>Crop Water Needs</i>	Indicator of water resource conservation capacity (mm)		X
<i>Water Use Autonomy</i>	Qualitative estimation based on water supply type (rainfall, catchment basin, river, aquifer)		X
<i>Energy Consumption</i>	I <sub>Energy</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>Energetic Efficiency</i>	ENE = OE/IE × 100	ENE: Energetic Efficiency (%) OE: Output Energy (GJ) IE: Input Energy (GJ)	X
<i>Crop Phosphorus Needs</i>	Indicator of phosphorus resource conservation capacity (kg of P)		X
<i>Phosphorus Use Autonomy</i>	AMP = (1- MP/TP) × 100	AMP: Autonomy with respect to Mineral Phosphorus (%) MP: Mineral Phosphorus (kg) TP: Total Phosphorus (kg)	X
<i>Crop Diversity</i>	I <sub>Div</sub>	Indigo Method (Qualitative/Ordinal score)	d
<i>Sprayed Area</i>	PSA = (1- NTC/TCC) × 100	PSA: Proportion of Sprayed Area (%) NTC: Number of non-Treated Crop Cycles TCC: Number of Total Crop Cycles	X
<i>Insecticides</i>	TFI <sub>Insecticides</sub>	TFI calculated for insecticides	f
<i>Fungicides</i>	TFI <sub>Fungicides</sub>	TFI calculated for fungicides	f
<i>Herbicides</i>	TFI <sub>Herbicides</sub>	TFI calculated for herbicides	f



appreciation, using three- to five-value qualitative scales defined by the workgroup on the basis of expertise and depending on the resolution of the attribute (Fig. 3A). Typically, the linguistic scales take the form of a ‘Low/Medium/High’ progression, with the addition of ‘Very Low’ or ‘Very High’ in some cases, depending on the attribute (gray rectangles in Fig. 3A). The transformation process may be of one of two types, as a function of the initial type of transformation: (i) from quantitative to linguistic, (ii) from ordinal (qualitative) to linguistic. When the initial information is linguistic in nature, no transformation is needed (e.g., *Physical Constraints, Compaction Risk, Water Use Autonomy*, Tab. I, Fig. 3A). Suggestions as to the correspondence between initial information and linguistic scales are made in some cases, especially for some I-Phy/INDIGO indicators (e.g., *Surface Water, Ground Water, Phosphorus Losses and Pesticide Emissions*), which initially display ordinal outputs that must first be interpreted according to the score grid provided by the INDIGO framework (e.g., 0 to 10 for I-Phy, Tab. I), before transformation into linguistic information. In other cases, as for economic attributes (e.g., *Profitability, Independence and Efficiency*), interpretation of the quantitative results and their transformation into linguistic assertions are left to the decision-maker’s discretion, because the interpretation process in these cases depends more on local/specific considerations and norms.

#### 2.4.2. The decision tree

The MASC decision tree (Fig. 3B) is a hierarchical structure aggregating 32 input attributes through 22 aggregate attributes, based on ‘If-Then’ decision rules defining aggregation rules expressed as ‘weightings’ (details in Sect. 2.2 and Fig. 2), for assessment of the overall sustainability of cropping systems (*Overall Sustainability* attribute). The hierarchical structure shown in Figure 3B was developed from expert knowledge (according to the working process described in Sect. 2.3) and illustrates the decision-modeling process underlying the MASC model. As for input attributes, the scales for aggregate attributes are linguistic and defined on the basis of expert knowledge, with a resolution of three to five units (gray rectangles below aggregate attributes in Fig. 3). The aggregation process in MASC may be free or predetermined.

**Free aggregation** occurs in  $\approx 70\%$  of the cases in MASC (thick gray branches in Fig. 3B). In these cases, the model does not specify the strict weighting values to be assigned to the attributes to be aggregated (numbers in gray boxes refer to aggregation rules specific to the evaluation presented in Sect. 2.5). Instead, the choice is left open to the decision-maker, so the aggregation of attributes depends on:

- (i) Personal strategic views (e.g., aggregation of economic attributes or certain social attributes);
- (ii) Local and specific environmental contexts and norms (e.g., aggregation within *Water Pollution Risks* sub-model);
- (iii) The decision-makers’ broad representations of the sustainability assessment problem for their cropping systems, defining the priorities between (e.g., aggregation of

economic, social and environmental sustainability) and within sustainability dimensions.

However, depending on the specific socioeconomic and environmental context of the cropping system to be assessed, it may be relevant to provide the decision-maker with expertise-based thresholds for the weightings to be used in the aggregation process. An example of this situation is provided in Section 2.5.

**Predetermined aggregation** is typical of the aggregations developed from expert knowledge by the workgroup, and may be considered valid regardless of the context and cropping system (Fig. 3B, thick black lines and numbers in black boxes). These aggregations define submodels (e.g. *Water Conservation* and *Biodiversity Conservation* submodels), which can be interpreted as ‘synthetic indicators’ processing input attributes and linking them with the decision tree.

#### 2.5. Specific settings for the first test of MASC

Input information processing for these systems followed the steps described in Section 2.4.1. The aggregation rules followed for this specific assessment (numbers in gray boxes in Fig. 3B) were generated through interactions between the eight decision-makers/analysts of the workgroup.

The values used are offset from threshold aggregation rules, which were defined by the workgroup for most aggregations, by 15 to 33%. In the absence of sensitivity analysis, the use of these settings ensured that the sustainability evaluation process was not distorted or biased by excessive weighting values (e.g. completely abolishing a sustainability component, using a weighting of 0 for one of the three main dimensions). These threshold values were established from expert knowledge, based on:

- (i) Local considerations relating to the cropping systems (e.g. given the importance of soil quality problems in Picardy, minimum thresholds for *Physical Quality* and *Chemical Quality* attributes were set at 33% in the decision model) and
- (ii) Considerations relating to the importance of different issues underlying sustainability in general (e.g. given the importance of the issues relating to the *Environmental Quality, Abiotic Resources Conservation* and *Biodiversity Conservation* attributes, the minimum threshold value for aggregation was set at 20% for each of these attributes). A similar rationale could be applied to the aggregation of economic, social and environmental sustainability submodels, for which minimum aggregation thresholds were set at 15%.

### 3. RESULTS AND DISCUSSION

#### 3.1. First test of MASC: implication for future real-case studies

The coherence of the MASC model was tested by assessing four cropping systems generated from expert knowledge

(see Sect. 2.3), through the specific steps described above. The initial qualitative values of the input attributes, together with those relating to all the aggregate attributes resulting from combinations of these initial values, are given in Table II, with the aggregation rules followed for this assessment reported in Figure 3B (numbers in gray and black boxes).

A detailed interpretation of the results obtained would not be relevant in this study because the cropping systems used here were developed for testing the model, rather than for a real comparison of sustainability. Nonetheless, the evaluation results were consistent with the initial representation of the cropping systems: as expected, the prevailing cropping system (CS#1) displayed high economic sustainability and low environmental sustainability, and the 3 alternative cropping systems developed on the basis of expertise (CS#2, CS#3 and CS#4, Fig. 4) had higher values for environmental and overall sustainability, ranging from ‘Medium’ to ‘Very High’. The decision rules used did not generate ‘illogical’ outputs, at least for this test, which provides a first approximation of the coherence of the logic behind the disaggregation-aggregation approach in MASC. This outcome paves the way for a more formal assessment of realistic cases.

### 3.2. Relevance of the holistic representation of the sustainability concept at the cropping system scale

MASC explicitly and specifically functions at the scale of the cropping system, and this is an important characteristic. The cropping system scale has a granularity intermediate between those of the single plot/single year and farming system scales. The use of this scale makes it possible to isolate the effects on overall socioeconomic and environmental performance of different cropping activities within a farming system, and to take interactions between individual crops in the process into account in an explicit manner.

A second key feature of MASC is that it explicitly formalizes a global concept of sustainability, *discretized* and encapsulated into a vector of 32 input criteria (Fig. 3, Tab. I). Although MASC addresses the relatively limited scale of the cropping system, a critical objective of this study was to make this model as holistic as reasonably possible, by addressing issues not classically addressed at this scale but considered by a growing number of researchers to be ‘challenging targets for future agriculture’ (see Kirchmann and Thorvaldsson, 2000 for review). MASC may therefore be considered a “stand-alone” tool, which is not the case for other models analyzed in similar studies assessing cropping systems. Indeed, comparable studies have either targeted (i) specific objectives underlying sustainability, such as the impact of cropping systems on soil quality (Bohanec et al., 2007), and the economic and ecological impacts of genetically modified crops (Bt maize; Bohanec et al., 2008) or (ii) overall sustainability, but through a limited number of sustainability criteria, as in the study by Mazzetto and Bonera (2003), who considered 9 criteria encompassing economic, environmental and technical issues to

be aggregated within a specifically designed sustainability assessment tool.

Classically, studies addressing sustainability assessment at larger scales, such as farming systems, benefit from a larger number of input criteria (e.g., frameworks of Phillis and Andriantiatsaholiniainan, 2001; Häni et al., 2003; Van Cauwenbergh et al., 2007; Meul et al., 2008). Thus, for the specific needs of the MASC, we either (i) directly ‘imported’ criteria generally considered at higher levels or (ii) simply generated our specific criteria as a function of the issues targeted, based on a combination of literature review and expert knowledge.

Criteria were directly imported, for example, for certain economic sustainability factors considered by MASC. This was the case for the set of attributes reflecting economic independence and efficiency, which were imported from the IDEA framework (Vilain et al., 2008), to reflect medium-term economic sustainability objectives (in addition to short-term profitability, Fig. 3, Tab. I).

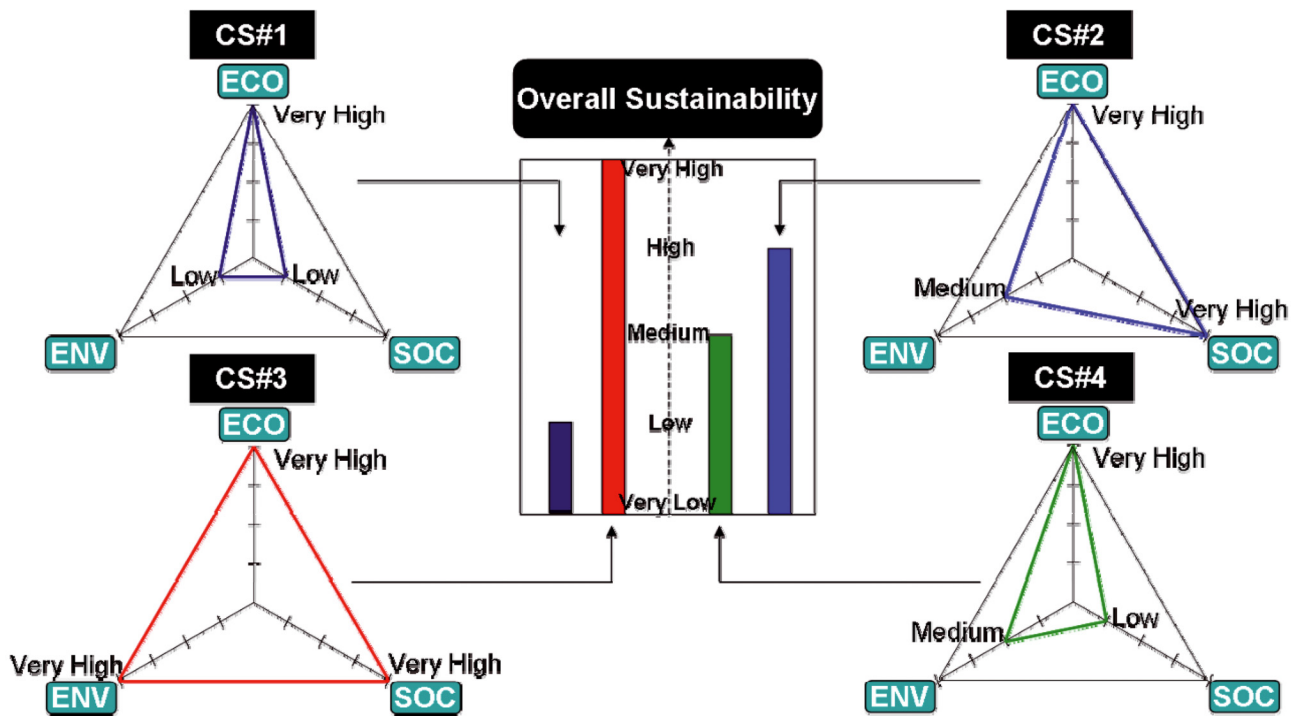
Specific criteria were generated in cases in which no ready-to-use indicators were available at cropping-system scale. This was the case for the ‘synthetic indicators’ of MASC (see Sect. 2.4.2), represented by the submodels aggregated on the basis of expert knowledge for both social (*Complexity of Implementation*) and environmental (*Water Conservation*, *Phosphorus Conservation* and *Biodiversity Conservation*) sustainability dimensions.

Classically, the social sustainability dimension is either not represented (Bohanec et al., 2008) or only implicitly considered in studies targeting cropping systems (e.g., ‘operating difficulty’ criterion as a part of a *technical evaluation*, Mazzetto and Bonera, 2003). In MASC, we chose to address this issue more explicitly, by developing an indicator dealing with the *Complexity of implementation* in the context of the worker’s quality of life and, thus, as a component of the social sustainability of the cropping system. Furthermore, in line with the recommendations of Calker et al. (2007), we explicitly considered the health of agricultural workers in our assessment. Two complementary standpoints, each represented by a sustainability criterion, were used for this analysis: the actual constraints experienced by the worker, based on an interpretation grid provided by the MSA (*Mutualité Sociale Agricole*, 2006, Tab. I) and risks due to contact with hazardous substances.

For the *Water Conservation* attribute, we focused on water consumption as a function of local critical periods and the type of water supply, rather than on absolute water consumption values or water use efficiency-based factors, as the principal aim was to estimate the local impact of the cropping system on the availability of water as an increasingly rare (and shared) resource. Similar approaches were followed in the RISE (Häni et al., 2003) and MOTIFS (Meul et al., 2008) sustainability assessment frameworks, although neither of these models explicitly took into account the occurrence of critical periods of water shortage, which are becoming increasingly frequent in many temperate areas of Europe, and elsewhere. The same type of reasoning was applied to *Phosphorus Conservation*. Phosphorus use is taken into account in many sustainability assessment frameworks, mostly in terms of losses, use efficiency

**Table II.** Detailed results of the sustainability assessment carried out with MASC for four cropping systems generated from expert knowledge. <sup>a</sup> Compacted hierarchical representation of the decision tree shown in Figure 3B. The root attributes of the whole model (*Overall Sustainability*) and those of the Economic, Social and Environmental submodels are represented with black and gray backgrounds, respectively. Aggregate attributes are presented in bold typeface. The number of points preceding the name of each attribute indicates its hierarchical distance from the root attribute. <sup>b</sup> Qualitative values of all the attributes of the MASC model, for each assessment. The scales from which the qualitative values of each attribute were taken are reported in Figure 3 (gray rectangles to the right of and below input and aggregate attributes, respectively). Bold qualitative values are those resulting from aggregation, according to the values of descendants and the predefined aggregation rules (numbers in gray and black boxes in Figure 3B). <sup>b1</sup> Cropping system (CS), inspired from actual systems in Picardy (northern France), described in the “ADAR Systèmes de cultures innovants” project. <sup>b2</sup> Three alternative systems described in the “ADAR Systèmes de culture innovants” project.

Attributes Hierarchy <sup>a</sup>	Attribute values and Assessment results <sup>b</sup>			
	Actual <sup>b1</sup>	Alternatives <sup>b2</sup>		
	CS#1	CS#2	CS#3	CS#4
<b>Overall Sustainability</b>	<b>Low</b>	<b>High</b>	<b>Very High</b>	<b>Medium</b>
<b>. Economic Sustainability</b>	<b>Very High</b>	<b>Very High</b>	<b>Very High</b>	<b>Very High</b>
.. Profitability	Medium	High	High	High
<b>.. Autonomy</b>	<b>Medium</b>	<b>Very High</b>	<b>Medium</b>	<b>Medium</b>
... Independence	Medium	High	Medium	Medium
... Efficiency	Medium	High	Medium	Medium
.. Specific Equipment Needs	Low	Medium	Medium	Low
<b>. Social Sustainability</b>	<b>Low</b>	<b>Very High</b>	<b>Very High</b>	<b>Low</b>
.. Contribution To Local Employment	Medium	Very High	Very High	Medium
<b>.. Operational Difficulties</b>	<b>Medium</b>	<b>Low</b>	<b>Low</b>	<b>Low</b>
... Physical Constraints	High	Low	Medium	Medium
<b>... Complexity Of Implementation</b>	<b>Low</b>	<b>High</b>	<b>High</b>	<b>Medium</b>
.... Number Of Crops	Low	High	High	Medium
.... Number of Specific Operations	Low	Medium	Medium	Medium
.. Health Risks	High	Very Low	Low	Medium
<b>. Environmental Sustainability</b>	<b>Low</b>	<b>Medium</b>	<b>Very High</b>	<b>Medium</b>
<b>.. Environmental Quality</b>	<b>Very Low</b>	<b>Low</b>	<b>Medium</b>	<b>Low</b>
<b>... Water Pollution Risks</b>	<b>Medium</b>	<b>Low</b>	<b>Low</b>	<b>Medium</b>
.... <b>Pesticide Losses</b>	<b>Low</b>	<b>Very Low</b>	<b>Low</b>	<b>Low</b>
..... Surface Water	Low	Low	Low	Low
..... Ground Water	Medium	Low	Medium	Medium
.... NO3 Losses	Low	Low	Very Low	Medium
.... Phosphorus Losses	Medium	Medium	Low	Medium
<b>... Air Pollution Risks</b>	<b>Medium</b>	<b>Low</b>	<b>Low</b>	<b>Low</b>
.... NH3 Emissions	Very High	Low	Medium	Low
.... N2O Emissions	Medium	Medium	Medium	Medium
.... Pesticide Emissions	Medium	Low	Low	Medium
<b>... Soil Quality</b>	<b>Low</b>	<b>Low</b>	<b>Medium</b>	<b>High</b>
.... <b>Physical Quality</b>	<b>Medium</b>	<b>Medium</b>	<b>Medium</b>	<b>High</b>
..... Compaction Risk	Low	Medium	Low	Medium
..... Erosion Risk	Medium	Medium	Medium	Low
.... <b>Chemical Quality</b>	<b>Low</b>	<b>Low</b>	<b>Medium</b>	<b>High</b>
..... Organic Matter Content	Medium	Medium	Medium	Medium
..... Phosphorus Fertility	Very Low	Very Low	Low	Medium
<b>.. Abiotic Resources Conservation</b>	<b>Low</b>	<b>Medium</b>	<b>Medium</b>	<b>Medium</b>
<b>... Water Conservation</b>	<b>Medium</b>	<b>Medium</b>	<b>Low</b>	<b>Medium</b>
.... Dry Period Irrigation Needs	<b>Medium</b>	<b>Medium</b>	<b>High</b>	<b>Medium</b>
<b>.... Dependency On Water</b>	<b>Medium</b>	<b>Medium</b>	<b>Medium</b>	<b>Medium</b>
..... Crop Water Needs	Medium	Medium	High	Medium
..... Water Use Autonomy	Medium	Medium	High	Medium
<b>... Energy Conservation</b>	<b>Low</b>	<b>High</b>	<b>High</b>	<b>High</b>
.... Energy Consumption	Medium	Low	Low	Low
.... Energetic Efficiency	Low	High	High	Medium
<b>... Phosphorus Conservation</b>	<b>High</b>	<b>High</b>	<b>High</b>	<b>High</b>
.... Crop Phosphorus Needs	Medium	Medium	Medium	Medium
.... Phosphorus Use Autonomy	High	High	High	High
<b>.. Biodiversity Conservation</b>	<b>Very Low</b>	<b>High</b>	<b>High</b>	<b>Medium</b>
... Crop Diversity	Low	High	High	Medium
<b>... Pesticide Use Intensity</b>	<b>Very High</b>	<b>Low</b>	<b>Low</b>	<b>Medium</b>
.... Sprayed Area	High	Very Low	Low	Medium
<b>.... Number Of Doses</b>	<b>High</b>	<b>Very Low</b>	<b>Very Low</b>	<b>Medium</b>
..... Insecticides	Medium	Low	Low	Medium
..... Fungicides	Medium	Low	Medium	Low
..... Herbicides	High	Low	Low	Medium



**Figure 4.** Graphical representation of the results of the sustainability assessment carried out in MASC for four cropping systems generated from expert knowledge. Radar plots represent the evaluation results for each of the three sustainability dimensions considered by the model, where ECO, SOC and ENV refer to the economic, social and environmental sustainability assessment submodels, respectively. CS#1 to CS#4 refer to cropping systems #1 to #4, respectively. The overall sustainability evaluation results for each cropping system are reported in the central bar graph, with the qualitative scales inserted in the middle. Each color refers to the corresponding cropping system. Details of the (i) qualitative values of both input and aggregate attributes and (ii) aggregation rules corresponding to these results are reported in Table II and Figure 4, respectively.

or soil fertility (Häni et al., 2003; Dogliotti et al., 2004; Meul et al., 2008), but few, if any, of these studies considered phosphorus use explicitly from this ‘rarefaction’ standpoint, so the importation of pre-existing indicators was not successful.

For *Biodiversity Conservation*, previous studies have reported a plethora of biodiversity indicators for measuring the impact of agricultural systems on surrounding ecological niches, but these indicators are designed for use with much larger scales, such as landscapes and regions (Osinski et al., 2003; Hietala-Koivu et al., 2004). Many of these indicators are actually ‘bioindicators’ measuring the impact on population dynamic parameters of specific groups or taxa of animals (mainly insects) and plants (Paoletti, 1999; Döring et al., 2003; Parisi et al., 2005). We felt that it was irrelevant to approach biodiversity assessment ex ante and at the limited scale of the cropping system from the bioindicator standpoint, because (i) this approach may be cumbersome (Braband et al., 2003) and (ii) a given bioindicator may be valid in one context but not in another. We therefore used an approach based on indirect estimation of the impact of the cropping system on biodiversity conservation, considering *Pesticide Use Intensity* to be representative of a ‘pressure’ component and *Crop Diversity*, which is assumed to be positively correlated with biodiversity conservation in many assessment frameworks (see Braband et al., 2003 for review).

Overall, the resulting vector of 32 sustainability criteria may be considered to give an unbalanced representation of the three dimensions of sustainability, in that economic and social criteria are outnumbered by environmental criteria. However, this is not specific to the MASC model, as a similar ‘imbalance’ is observed in most sustainability assessment-dedicated frameworks, functioning at different levels of agricultural systems. This situation may result from the monetary nature of economic criteria, rendering these criteria less holistic than environmental criteria, which cover a large variety of complex and intricate problems. Social criteria occupy an intermediate position, as more considerations, such as landscape quality, animal well-being, food security and safety and community function, may be taken into account, depending on the scale at which the assessment is carried out.

### 3.3. Sustainability assessment modeling by MASC: advantages and restrictions

#### 3.3.1. Advantages

##### 3.3.1.1. Underlying MCDA methodology

A functional key feature of MASC lies in its underlying MCDA methodology. This methodology was selected in

accordance with the recommendations we issued in a previous paper (Sadok et al., 2008) for identifying relevant tools for the ex ante assessment of cropping systems. The functional value of MASC lies partly in its implementation within a DSS (decision support system) using a ‘mixed’ (or non-classical) MCDA method for the explicit handling of qualitative information and the expression of preferential information through qualitative decision rules aggregating initial criteria. Such features are not common within the large group of MCDA tools used for sustainability assessment of agricultural systems (i.e., Häni et al., 2003; Dogliotti et al., 2005; Meyer-Aurich, 2005; Van Cauwenbergh et al., 2007; Meul et al., 2008). Explicitly qualitative information can capture the uncertainties inherent to sustainability assessment (especially in the ex ante mode) and more realistically integrate the decision-maker’s own views, as they are not necessarily expressed through formal, quantitative models (Dent et al., 1995; Munda, 2005). In addition, the method used here handles incomparability, incommensurability and non compensation between sustainability dimensions (see review by Sadok et al., 2008) more efficiently than other more ‘classical’ MCDA approaches frequently used for sustainability assessment, such as ELECTRE (Loyce et al., 2002; Mazzetto and Bonera, 2003), AHP (Shrestha et al., 2004) or multi-objective optimization tools (Dogliotti et al., 2005; Meyer-Aurich, 2005).

### 3.3.1.2. Representation of the sustainability assessment problem

In terms of decision modeling, one of the specific features of MASC is the way in which it conceptualizes the sustainability assessment problem (Fig. 3). During the development of this model, it was decided that this representation should not follow a cause-effect modeling approach (i.e., relationships between the properties of the cropping system and their impacts). The representation of these relationships would have been highly cumbersome, given the high level of interdependency between the economic, social and environmental dimensions. We therefore used a *decisional* approach to break the assessment problem down. This approach was based on an interpretation of a frequently used definition of ‘sustainability’ (three dimensions, each representing a hierarchy of sustainability objectives), making the assessment much more manageable. This approach has two major advantages.

The first is an ‘operational’ advantage, resulting from the compromise between the maintenance of realistic representations of the sustainability assessment decisional problem and limiting the number of decision rules implemented. This compromise keeps the decisional problem to a manageable size, as recommended by Ma (2006). It resulted in the MASC model having far fewer decision rules than the model used by Bohanec et al. (2008), which addressed the specific problem of the ecological and economic assessment of cropping systems based on genetically modified crops and developed on the basis of the same MCDA methodology.

The second advantage is the potential ‘pedagogical’ value of MASC. Pedagogical value is not generally considered

essential in sustainability assessment frameworks, in which preference models are partly or totally opaque to the users. The software version of MASC (MASC\_V1.0, implemented within the DSS tool DEXi) provides a graphical and dynamic representation of the whole hierarchy, from the input criteria to the root attribute, making it possible for users to trace the effects of changes in the value of a single criterion on the overall result of the assessment. This feature is particularly relevant in cases in which the decision-maker is unaware of many of the implications of sustainability at the cropping system scale, which is more often the case for non specialists. This pedagogical value is enhanced by the need for users to specify the weightings to be applied to different dimensions and the underlying objectives, as a function of their own contexts and priorities, requiring them to supply their own definition of sustainability. In addition, the use of qualitative information at all levels of the model (input attributes, *If-Then* decision rules) probably makes this model more accessible to a wider range of decision-makers than more mechanistic, quantitative models, as qualitative models are able to represent more realistically empirical knowledge resulting from personal views and experiences.

### 3.3.2. Restrictions on the use of MASC

#### 3.3.2.1. Qualitative modeling limitations

The exclusive use of qualitative input information by MASC inevitably limits its resolution, as quantitative data cannot be used directly. Moreover, the use of qualitative data requires greater attention in the description of the hypotheses adopted and of the procedure used to develop criteria, because qualitative data are more difficult to interpret objectively than quantitative data (Andreoli et al., 1999). This is particularly the case for the transformation into qualitative (linguistic) values for certain input quantitative attributes of MASC. There is a risk that the link between the ‘absolute’ quantitative value and the ‘chosen’ qualitative scale is not unique and depends on the decision-maker. However, this situation may also render the tool more flexible, as, for example, a semi-net margin of 700 €/ha may be translated into different qualitative values (e.g., ‘Low’, ‘Medium’, ‘High’) depending on the decision-maker’s opinion and the context. A balance between limitation and flexibility must therefore be reached. MASC (MASC\_V1.0) software includes a tutorial, with guidelines for the quantitative-qualitative transformation for each input attribute.

In addition, the fact that the models implemented with the DEXi DSS are based on ‘*If-Then*’ decision rules — which can define only relationships between conditions and consequences — hampers the modeling of the intrinsic properties and functioning of the systems considered (Bohanec et al., 2008). This makes it difficult to investigate the intrinsic sources of differences between the cropping systems assessed, unless the approach is combined with quantitative, mechanistic models (e.g., crop models, management plans, etc.). This was not a critical limitation in this study, because MASC aims

to compare cropping systems on the basis of their impact on sustainability, rather than to identify the sources of differences between the alternatives. It would probably be more relevant to consider this aspect at the cropping system generation phase of the framework.

### 3.3.2.2. Sustainability assessment limitations

Assessments of holistic entities, such as sustainability — particularly from an ex ante perspective — inevitably result in significant uncertainty. According to Walter (2006), this uncertainty is threefold: (i) social and technological, because needs and priorities are continually changing, (ii) epistemic, due to inevitable gaps in the knowledge representing economic, social and environmental systems and (iii) stochastic, due to random, unpredictable phenomena in both natural (e.g., climatic) and social systems (e.g., economic crisis).

Since the early stages of development of the MASC model, it has been clear that one of the principal weaknesses of this model lies in its not taking these uncertainties into account in an explicit manner. We feel that this criticism could also be leveled at most sustainability assessment frameworks. However, MASC provides opportunities to handle the first two types of uncertainty, at least to some extent, through (i) the flexibility it offers the user in terms of the choice of aggregation rules (70% of which are left open to the decision-maker) and (ii) the use of qualitative modeling to deal with the inevitable lack of quantitative knowledge for some issues underlying sustainability assessment for cropping systems. For stochastic uncertainties, the sensitivity of the sustainability of a cropping system to a ‘stochastic’ event could be assessed by including this variation in the calculation/estimation of all criteria rather than in the decision model itself. This would make it possible to compare different assessment results for the same cropping system subjected to different sets of environmental/socioeconomic conditions. In this context, it might be possible to connect the formalism underlying the input attributes to the databases generating scenarios. An approach of this kind is already planned in the framework of sensitivity analysis for the decision model.

## 4. CONCLUSION

The model described here is the first version of a registered tool, *MASC\_V1.0*, a key element of an ex ante (i.e., ‘*Before-the-event*’) assessment framework for use at cropping system level. Such tools are designed to assess the sustainability of different cropping systems *before* their implementation in the field. This approach is potentially of prime importance in a world in which the economic, social and environmental constraints and objectives shaping sustainability change so rapidly that the window for classical in-field prototyping approaches is very small.

The key features of this tool are its (i) scale of assessment; it explicitly considers the cropping system level, which is viewed as the basic functional unit making up a farming system, (ii)

unique vector of 32 input criteria across the three dimensions of sustainability spanning a wide spectrum of sustainability objectives, (iii) the use of a ‘mixed’, alternative MCDA method able to handle qualitative information and to express preferential information through qualitative *If-Then* decision rules, which are easy to handle by non specialists and (iv) its transparent and dynamic decision model, providing a formal definition of sustainability helping decision-makers to understand the implications of their views, experiences and decisions for the overall sustainability of their system. The implementation test presented here should be seen as the first of a series of tests, culminating in the production of version 1.1 of the tool.

A detailed analysis of the sensitivity of MASC is already underway. This study should make it possible to estimate the granularity offered by this tool for different decisional (i.e., aggregation rules) and cropping system representation (i.e., input attribute values) scenarios, taking into account pedoclimatic and socioeconomic fluctuations. This work may lead to adjustments to the definition of scales and aggregation rules in MASC.

Another short-term perspective concerns testing of the generic nature of the model for a large range of cropping systems and types of decision-makers. Such testing is also already underway because MASC is currently available to various groups. We expect it to be necessary to make some adjustments to the model, the most significant of these changes occurring at the input attribute level. However, the current version of MASC is flexible as to the choice of formalism underlying its input attributes, as we feel it is more appropriate to allow the replacement of certain formalisms by others more in line with the decision-maker’s own views, context or cropping system type. This flexibility has already been used in the development of a variant of the MASC model for the assessment of dairy grazing systems (Coquil et al., 2007).

In the medium term, we plan to use a more formalized approach for generating the cropping systems to be assessed ex ante. This requires the development of an explicit generation module based on simulation tools (Bergez et al., 2008) or combinatorial approaches (Dogliotti et al., 2003). We are also considering the possibility of developing an expert system in the DSS in which MASC is implemented, as a tool for generating cropping systems.

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